



Unsupervised Visual Representation Learning by Graph-Based Consistent Constraints

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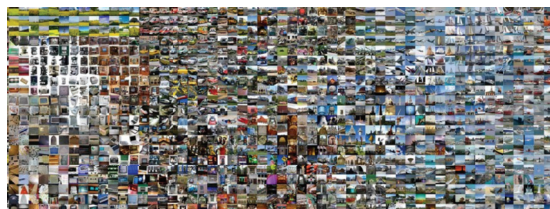
ECCV'16
 EUROPEAN CONFERENCE
 ON COMPUTER VISION

<http://bit.ly/feature-learning-eccv2016>

Introduction

Supervised learning: Expensive annotations & poor scalability

Goal: Visual representation learning with a large, unlabeled image collection



Prior work:

Context [Doersch et al. ICCV'15]

Tracking [Wang and Gupta ICCV'15]

instance-level training data **within** the same image/video

Contributions

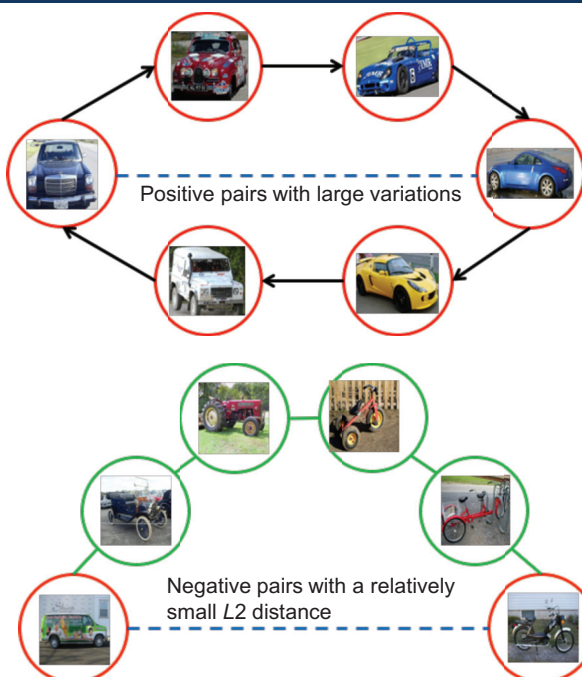
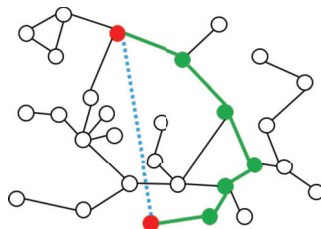
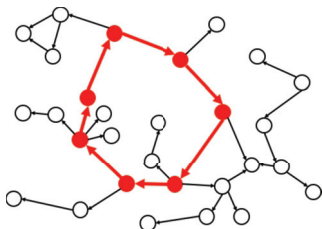
- Generating semantically similar and dissimilar image pairs from a large, unlabeled image collection via an unsupervised constraint mining approach.
- Learning visual representations in an unsupervised manner via binary classification based on Siamese network.
- Improving the classification performance by applying the mined constraints for both unsupervised feature learning and semi-supervised learning.

Unsupervised Visual Representation Learning

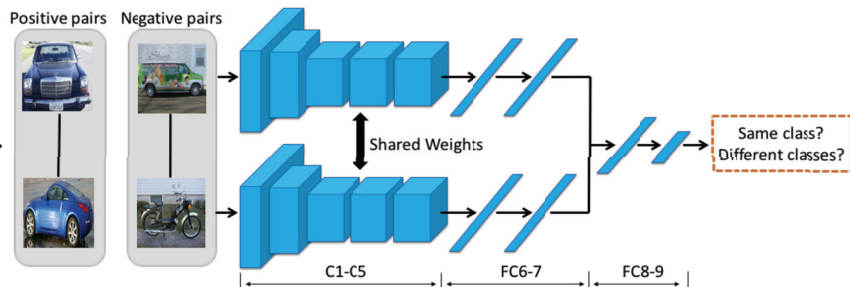
Mining: Generate **category-level** training samples **across** different images

Cycle consistency → Positive mining

Geodesic distance → Negative mining



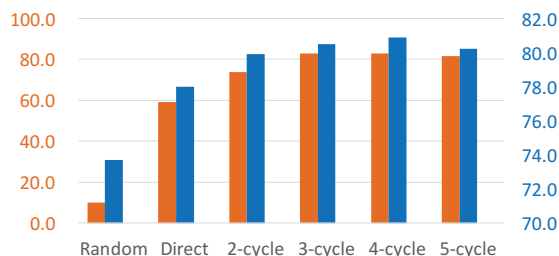
Training: Learn visual representations for **binary** classification



Experiments

Positive mining (CIFAR10)

TP rate Accuracy



Accurate positive pairs & Better CNN representations

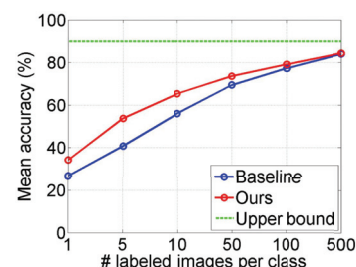
Image classification (VOC'07)

Methods	Supervision	mAP
Agrawal ICCV'15	Ego-motion	52.9
Doersch ICCV'15	Context	55.3
Wang ICCV'15	Tracking triplet	58.4
SIFT+FV	Matching pair	46.0
Ours	Matching pair	56.5
Krizhevsky NIPS'12	Class labels	69.5

Competitive performance with the state-of-the-arts

Significant improvement over hand-crafted features

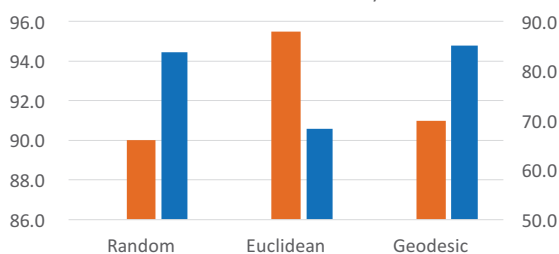
Semi-supervised learning (CIFAR10)



Boosted performance over directly fine-tuning

Negative mining (CIFAR10)

TN rate Accuracy



Euclidean
 Easy samples
 Geodesic
 Hard samples

Image search (ImageNet)



Comparable to supervised learned representations