

# CLEAR: Cumulative LEARning for One Shot One Class Image Recognition Jedrzej Kozerawski, Matthew Turk

## Objective

Learn to recognize novel object categories from only a single positive example and no negative examples.



### Motivation

Currently accurate object recognition requires many examples, both positive and negative.

People understand how to recognize a novel object based on:

- What this object is (and not what it is not)
- Overall lifetime experience (for context)

The question is: How can we mimic such human perception abilities?

### One Shot One Class

Data availability		Positives				
		Many	One	Zero		
Negatives	Many	Recognition	One Shot	Zero Shot		
	None	One Class	One Shot One Class	Zero Shot One Class		

One-Shot One-Class seeks to classify novel instances based on a single positive training example, with no explicit negative examples

Idea: Infer parameters of a weak classifier from a single image. How: Train DNN to do regression using pairs {image, classifier parameters}



**Test**: Create a classifier for a novel category using one, unseen image. Next use CNN as a feature extractor and classify query images using newly created classifier.



#### Proposed approach



#### Results

We have trained our model with the data from the ImageNet ILSVRC 2012 challenge and then tested it on five different datasets:

- Caltech 256
- Caltech-UCSD-Birds 200 2011
- Flowers 102
- MIT Indoor Scene Recognition 67
  - SUN attributes

The F1 score and MAP results for three of them are available below:

Method	Caltech-256		Flowers 102		MIT Indoor 67		
	F1	MAP	F1	MAP	F1	MAP	
hreshold	800.0	0.0004	0.0029	0.03	0.0001	0.001	
Chance	800.0	0.005	0.019	0.013	0.029	0.019	
-Class SVM	0.011	0.061	0.016	0.068	0.003	0.025	
EAR [ours]	0.176	0.364	0.087	0.193	0.111	0.209	

Larger values are better

One

#### Conclusions

This work addressed a novel problem of a One Shot **One Class classification** 

Our proposed approach shows that it is possible to learn weak classifiers from only a single positive example and no negative examples

Further work is necessary to overcome the class granularity problem (understand the breadth of a category from a single example)