

# Proofpoint Logo Detection & Classification

18th March 2021

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# Agenda

- Logistics
- Training Classifier with Corrected Aspect Ratio
- KNN Based Open Set Learning
- Open Long Tail Recognition

# Logistics

- Proofpoint Presentation on 4/22 at 12pm.
- Conclude exploratory work by 1st April (next Proofpoint meeting)
- Develop Unseen Test Set and Retrain models by 15th April.
- Submit deliverables to Proofpoint by 30th April.

# Main Deliverables to Proofpoint

1. Code used to run predictions.
2. Code used to train models.
3. Code use for error/performance analysis.
4. Model Artifacts.
5. All images and annotations.
6. Write Ups.

# Corrected Aspect Ratios

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# Augmentation Adjustments

- Resizing
  - Previously used standard resize in PyTorch
    - Did not maintain aspect ratio
- Color Jitter
  - Parameters were not set correctly → no impact upon inspect
  - Adjusted brightness, contrast, saturation, hue
- RandomResizeCrop and RandomRotation(90 degrees) maintained

# Previous Resizing



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# New Augmentation Examples



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# 1098 Classes Performance on Test Set

	Accuracy	Precision	Recall	F1-Score
Old Aug	0.9373	0.9780	0.9373	0.9532
New Aug	0.9451	0.9825	0.9451	0.9583

# Closed Set (with Unknown) Performance

	Accuracy	Precision	Recall	F1-Score
Old Aug	0.8950	0.9120	0.8950	0.8978
New Aug	0.9210	0.9291	0.9210	0.9224

# Open Set Learning

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# Terminology

**Correct Unknowns:** Percentage of true unknowns correctly classified as unknown.

**Incorrect Unknowns:** Percentage of known logos **incorrectly** called unknown.

**Accuracy Open Set:** Accuracy of model on known logos **given** that it was trained with the assumption that it would be deployed in the open world.

# Open Set Performance (32 Known Classes)

	Correct Unknowns	Incorrect Unknowns	Accuracy Open Set
<b>Unknown Class</b>	<b>84.4%</b>	<b>4.0%</b>	<b>94.5%</b>
<b>OLTR 1</b>	86.7%	16.4%	82.4%
<b>OLTR 2</b>	<b>88.9%</b>	<b>8.9%</b>	<b>90.7%</b>
<b>KNN (1st)</b>	92.2%	10.2%	89.5%
<b>KNN (2nd)</b>	93.3%	9.4%	90.4%
<b>KNN (1st + 2nd)</b>	<b>97.8%</b>	<b>9.4%</b>	<b>90.4%</b>

# Open Long Tail Recognition - OLTR

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# Large-Scale Long-Tailed Recognition in an Open World

## Large-Scale Long-Tailed Recognition in an Open World

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### Abstract

Real world data often have a long-tailed and open-ended distribution. A practical recognition system must classify among majority and minority classes, generalize from a few known instances, and acknowledge novelty upon a never seen instance. We define Open Long-Tailed Recognition (OLTR) as learning from such naturally distributed data and optimizing the classification accuracy over a balanced test set which include head, tail, and open classes.

OLTR must handle imbalanced classification, few-shot learning, and open-set recognition in one integrated algorithm, whereas existing classification approaches focus only on one aspect and deliver poorly over the entire class spectrum. The key challenges are how to share visual knowledge between head and tail classes and how to reduce confusion between tail and open classes.

We develop an integrated OLTR algorithm that maps an image to a feature space such that visual concepts can easily relate to each other based on a learned metric that respects the closed-world classification while acknowledging the novelty of the open world. Our so-called dynamic meta-embedding combines a direct image feature and an associ-

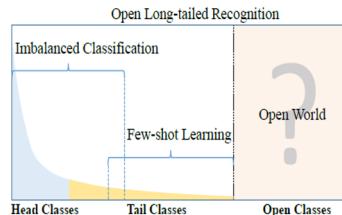


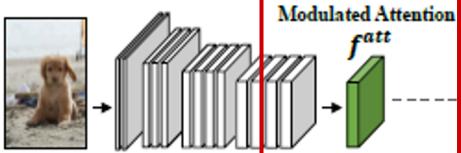
Figure 1: Our task of open long-tailed recognition must learn from long-tail distributed training data in an open world and deal with imbalance classification, few-shot learning, and open-set recognition over the entire spectrum.

While the natural data distribution contains head, tail, and open classes (Fig. 1), existing classification approaches focus mostly on the head [8, 30], the tail [55, 27], often in a closed setting [59, 34]. Traditional deep learning models are good at capturing the big data of head classes [26, 20]; more recently, few-shot learning methods have been developed for the small data of tail classes [52, 18].

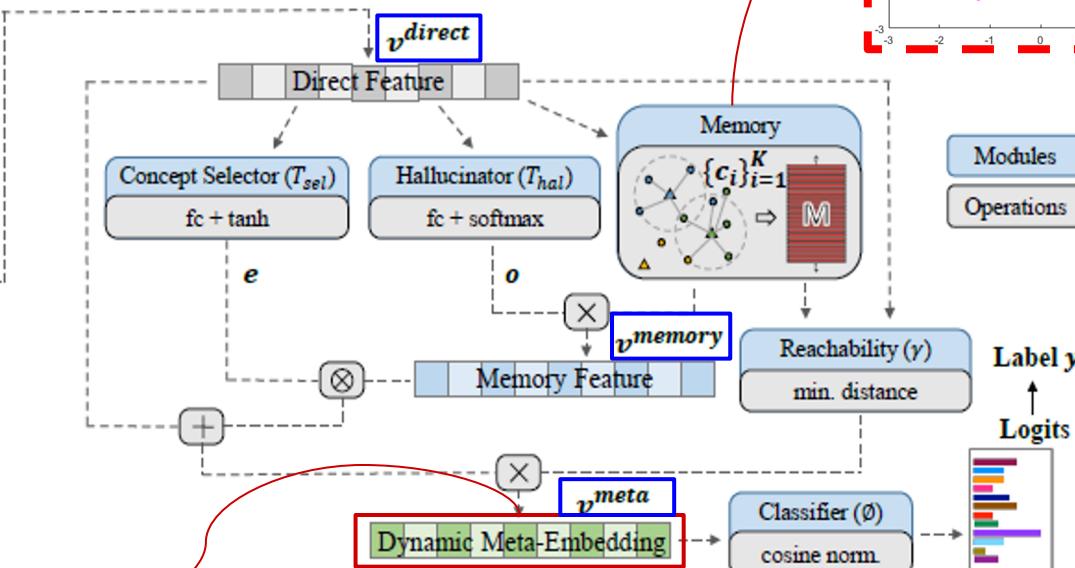
The screenshot shows a GitHub repository page for 'zhmiao/OpenLongTailRecognition-OLTR'. The repository has 32 stars, 604 forks, and 108 commits. The code tab is selected, showing a list of files and their commit history. The repository is described as a Pytorch implementation for "Large-Scale Long-Tailed Recognition in an Open World" (CVPR 2019 ORAL). The repository includes tags for computer-vision, deep-learning, open-set, long-tail, pytorch-implementation, cvpr2019, open-long-tail-recognition, otr, Readme, and BSD-3-Clause License. There are no releases or packages published.

Input Image

$x$



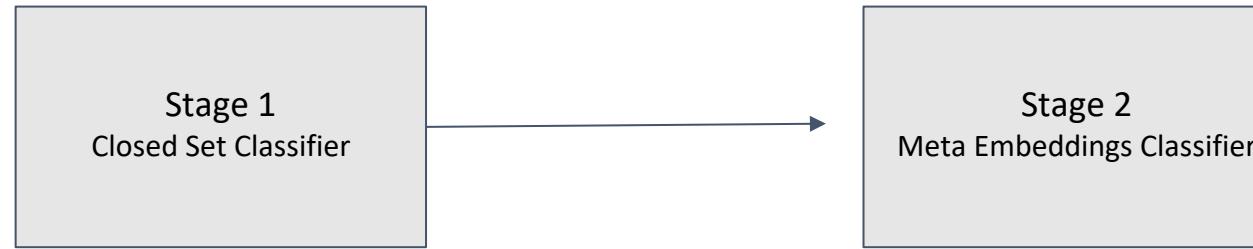
Modulated Attention  
 $f^att$



$$v^{meta} = (1/\gamma) \cdot (v^{direct} + e \otimes v^{memory})$$

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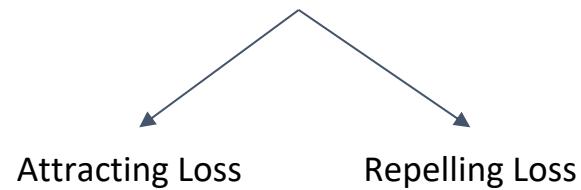
# Procedure



Categorical Cross Entropy Loss

Stage 2  
Meta Embeddings Classifier

Discriminative Centroid Loss



Attracting Loss

Repelling Loss

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# Pros & Cons

## Pros:

- Very quick during train and test time. Much faster than kNN
- No need to store and retrieve activation vectors

## Cons:

- Not so great code
- Switching out architectures was complicated

# KNN Based Open Set Learning

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# Procedure

1. Get second last activation vectors (AVs) of all train and validation observations.
2. Get distances of all true and unknown validation AVs from train AVs.
3. For each class, tune a distance threshold that maximizes accuracy for that class.
4. If no unknown is predicted into a given class:  
$$\text{threshold} = \max(\text{distances}) + \text{std}(\text{distances})$$
5. Use tuned thresholds for prediction on test observations and test unknowns.

# KNN Open Set Demo (1098 Known Classes)

Correct Unknowns: 73.36%

Incorrect Unknowns: 6.18%

Accuracy on Open Set: 91.42%

Demo:

<https://colab.research.google.com/drive/12jBFIQvDchfbIGxWrmb-UGRcjoq6L6pv?usp=sharing>

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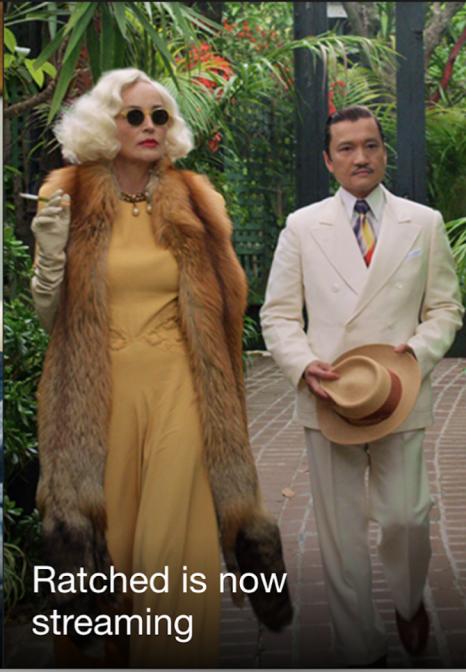
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How Monster Beverage Corporation is responding to the Coronavirus Outbreak: [LEARN MORE](#)

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26th May, 2020

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inequality in America

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# Drawbacks of KNN Open Set Learning

- Speed - need to calculate distances of each predicted logos to 12,000+ activation vectors.
- Storage
  - Activation vectors (AVs) need to be stored and loaded
  - Generate new AVs for slight changes in models; slow to do model testing
- Hyper parameter tuning - requires all known classes to have some incorrect predictions for threshold tuning.
- Threshold flexibility is limited: percentage of correct unknowns vs incorrect unknowns are challenging to adjust.

# Way Forward...

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