

Experiments Around Training Data Selection Methods for Image Classification

Department of Statistics Data Analysis Qualifying Exam

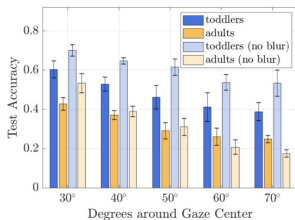
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Problem Statement and Objective

- ▶ Spurred on by both a wealth of image data and computational resources, deep convolutional neural networks are now widely used for image classification models
- ▶ CNNs may fit poorly when there is insufficient data.

Background: “Toddler-Inspired Visual Object Learning” (Crandall et al., 2018)

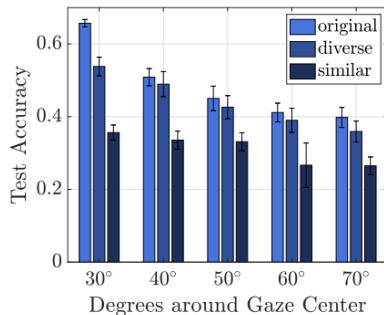
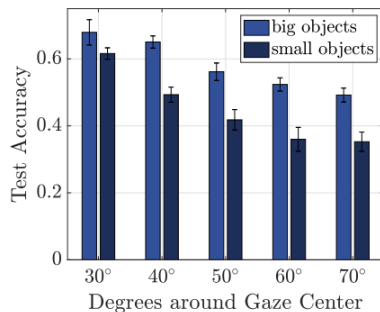
- ▶ Compared images taken by first-person cameras mounted on toddlers and parents
- ▶ Found that training VGG16 using toddler data resulted in higher test accuracy than training on parent data (same test set in both cases)



Background: “Toddler-Inspired Visual Object Learning” (Crandall et al., 2018)

- ▶ Distilled the differences in the datasets into two components: object size (how much of the image does the object take up) and image “diversity” (hard to measure)
- ▶ Subsampled the images to obtain training subsets of:
 - ▶ big objects
 - ▶ small objects
 - ▶ diverse images
 - ▶ similar images
 - ▶ random subset
- ▶ Image similarity/distance based on an image embedding method (GIST features)
- ▶ Cropped images to simulate multiple focal lengths
- ▶ Found that when objects are larger or when images are more diverse, test accuracy improves

Background: “Toddler-Inspired Visual Object Learning” (Crandall et al., 2018)



Outline and Summary

1. Replication studies
2. New methods for training data selection
3. Results
4. Conclusions and future work

Replication Study: Stanford Dogs dataset

Replication Study: CIFAR-10

Clustering