

CSCI E-81 Section 1: Machine Learning and Data Mining

Classifying Wildlife Camera Trap
Images for Conservation
International's Tropical Ecology
Assessment and Monitoring Network

Tony McGovern

Agenda

- ❖ Brief Literature Review
- ❖ Wildlife Image Classification Pipeline: Segmentation
- ❖ Wildlife Image Classification Pipeline: Feature Extraction
- ❖ Wildlife Image Classification Pipeline: Classification
- ❖ Wildlife Image Data: Analysis & Limitations
- ❖ Future Work

Problem Statement

- ❖ Conservation International's Tropical Ecology Assessment & Monitoring (TEAM) network has thousands of camera trap images of terrestrial mammals that for a variety of reasons require manual classification by experts in the field



Brief Literature Review

- ❖ Linear classifiers
 - ❖ Bolger et al. identified individual animals based on coat patterns for the analysis of mark recapture technique using SIFT keypoints and SVM classifiers
 - ❖ Yu et al. proposed a method for automated identification of animal species in camera trap images using dense SIFT descriptor and SVM classifiers
- ❖ Learning models
 - ❖ Goubin, et al. used graph cut with deep CNNs

Wildlife Image Classification Pipeline: Segmentation

- ❖ Chose to work with loxodonta species ~ 31K images
- ❖ Manually selected and cropped 287 images of the family
- ❖ Scaled the images down based on model run time



Wildlife Image Classification Pipeline: Feature Extraction

- ❖ OpenCV
 - ❖ Histogram Comparison
 - ❖ Template matching
 - ❖ Scale Invariant Feature Extraction
 - ❖ Speeded Up Robust Features - TBD
 - ❖ Local Binary Pattern (LBP) - TBD



Wildlife Image Classification Pipeline: Classification

- Linear Models
 - Most papers achieve good results with SVMs on small carefully segmented images using a variety of feature extraction methods
 - Bag of Visual Words employed for image mining with k-means clustering set to 10 cluster *a priori*
 - Accuracy using linear SVM with SIFT on 287 manually cropped training images and 574 test images got to 98%
- Learning Models
 - Attempted a CNN with dropout regularization between the fully connected layers with Keras/TF backend and achieved poor results

Wildlife Image Data: Analysis & Limitations

- ❖ Highly unbalanced classes of species
- ❖ Many of the images are misclassified
- ❖ There is a class of unlabeled images
- ❖ There is a class of unknown images
- ❖ Context matters: bursts of images are understood by experts as a sequence of images containing the identified species (part of animal species v whole animal species in frame problem)
- ❖ Existence of endangered species images preclude the use of citizen scientists to help classify images
- ❖ OpenCV offers many methods for image recognition but running them on many images is extremely slow

Future Work

- ❖ Increase training set with and use graph cut for segmentation
- ❖ Implement SURF and LBP texture descriptor as additional feature extraction methods
- ❖ Extend task to include other species images
- ❖ Short term goal is to help TEAM verify the classification of *elephantidae* family of elephants in their database
- ❖ Long term goal to automate classification from start to finish using deep learning for classification