Lion/Lioness Gender Recognition

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ABSTRACT
A Lion’s facial structure provides important features about its gender. This project presents an approach for gender detection from images considering facial structure as a region of interest. At the end of the project the system should be able to automatically detect the gender of a lion based on collected data. The project can potentially be expanded to cover a wider range of animals that can be used to keep record and prevent the extinction of protected species.

General Terms
Machine Vision, Classification problem, image processing

Keywords
Animal, face, lion, lioness, recognition, gender, species

1. INTRODUCTION
Modern software and hardware tools have allowed for advanced technique in facial recognition. What was formerly seen as science fiction only decades ago, are now realized. Today, there are some advanced facial recognition software that allows to compare and even track a person in real-time. These types of tools consider a facial structure and any type of elements that are unique and uses that database as query to track and identify similar faces. In this project, similar type of algorithm is used to identify a lion, whether it’s a male or female lion based on its unique facial feature.

There are some basic characteristics that allows us to distinguish between the male lion and its female counterpart. The scope of this project is limited to lions that have already aged. A male lion, in full age has a great amount of mane (facial hair) in comparison to a female lion. This feature can be used to differentiate them. A child to differentiate between the two genders of lions by showing a series of male and female lions and teach them based on difference between their features. Similarly, the program will be trained to identify and differentiate between a male and female lion. Two datasets with a series of male lions and female lions are used for comparison.

1.1 Creating the datasets

Figure 1. Female Lion
In the project file two subdirectories are created, each containing a set of images of 12 male lions and 12 female lions in the dataset accordingly. All the images are converted into grayscale and each image is resized to a 500x500 pixel image. This will serve as for our initial testing of our algorithm. See figure 1 and 2 for a montage of male and female lions.

Figure 2. Male Lion
1.2 Extraction of HoG Features
The algorithm is based on the method provided by MATLAB which is implemented for facial recognition on people. The histogram of oriented gradients (HOG) is a feature descriptor used in computer vision and image processing for object detection. The technique counts occurrences of gradient orientation in localized portions of an image.\cite{1}

![HoG Feature](image)

Figure 3. Extracted HoG Features from an image.

2. PROCEDURE
This method is similar to that of edge orientation histograms, scale invariant feature transform descriptors, and shape contexts, but differs in that it is computed on a dense grid of uniformly spaced cells and uses overlapping local contrast normalization for improved accuracy\cite{1}. HoG feature greatly differs between a male lion and a female lion because of mane. Figures 4 and 5 shows the extracted HoG features.

Data set is split into two as training and testing in 80:20 ratio. So, training set comprises of 20 images and testing set comprises of 4 images. HoG features are extracted for each image in data set and stored as training features. `extractHOGFeatures()` is the built-in function used to extract the HoG features. The returned features encode local shape information from regions within an image.

![Figure 4 Extracted HoG Feature of male lion](image)

Figure 4. Extracted HoG Feature of male lion

![Figure 5 Extracted HoG feature of female lion](image)

Figure 5. Extracted HoG feature of female lion

The features of the two classes are classified using a built-in function by MATLAB `fitcecoc` and stored in face classifier. `fitcecoc(Tbl,ResponseVarName)` returns a full, trained, multiclass, error-correcting output codes (ECOC) model using the predictors in table `Tbl` and the class labels in `Tbl.ResponseVarName`. `fitcecoc` uses $K(K - 1)/2$ binary support vector machine (SVM) models using the one-versus-one coding design, where $K$ is the number of unique class labels (levels).\cite{5,6,7} Similarly the HoG features are extracted for the testing data and later used as query to the built-in function `predict()`. `predict(Mdl,X)` returns a vector of predicted class labels (label) for the predictor data in the table or matrix $X$, based on the trained multiclass error-correcting output codes (ECOC) model.\cite{9,10}
3. RESULTS
Two images from each data set is used to test the program and results are accurate. If the extracted HoG features matches the respective class, it pulls the first image in the dataset as matched face. Figures 6 and 7 shows the results of testing set with a dataset of size 24.

3.1 Increasing the dataset
Dataset is increased to 40 images, 20 for the male and 20 for the female lions. Figure 8 and 9 shows a montage of the dataset with an increased size.
3.2 Results from increased dataset

Figure 10. Query of Lioness

Figure 11. Query of Male Lions

4. CONCLUSIONS
In conclusion, face detection in animals through image processing is an important and interesting challenge. Its application can be broadening to cover a variety of subjects and solve an unspeakable amount of problems where our imagination is the limit. For example, this project can be used in conducting a survey of animals in a biosphere reserve for the protection of endangered species. According to the world wild life, there are currently 18 different type of species on the critically endangered list, and 29 on the endangered list \[11\]. Just by having a visual of the animal we can keep record of animals and determine what species are becoming endangered and adding them to the protected list as needed.

5. REFERENCES