

Bird-O-Pedia An Automated Bird Classification System

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Abstract - Classifying a bird requires lots of knowledge, as Birds exhibit different features like color, beaks, wings, eye size, tail, sound, and live-in different inhabitants. Remembering species-specific information is not easy as there are around 9000 to 10,000 species. In this paper, we proposed a system backed by the CNN model which can identify a bird from an image and can provide information about that species. Our model is trained, validated, and tested on more than 28000 images of 200 species of birds. It model provided 97.44% accuracy on the training dataset and 98.40% on the testing dataset.

Key Words: Image classification, Convolutional Neural Network, Artificial Intelligence, Machine Learning.

1. INTRODUCTION

Birds can be found singing, chirping, and flying up and down from dawn to dusk, or activities like bird watching. Sometimes we can identify them or sometimes we aren't able to do so. This problem is not only faced by a normal person but also by an ornithologist. birds exhibit different colors, beaks, wings, eye sizes, tails, sounds, and live-in inhabitants. For this, we require a record book which contains the data of different birds. Searching for that is a hectic task. Many works have been done to identify birds but they were mainly based on their voice, as we know that collecting audio data is not only difficult but also contains much noise which results in erroneous results. So, we come up with an efficient method of Deep learning with CNN based web app which not only classifies a bird image but also provides basic information about that bird. The central objective of this system is to provide bird detail by just uploading images of that bird, we can also get information about a bird by using their name from the system in case we know bird's species. In this paper, the transfer learning [8] technique was used to retrain the InceptionV3 [7] model from Keras was used on the dataset of "two hundred" different bird species. The structure of the InceptionV3 [7] model which we used as a based model is described in section 3. Processes involved from data collection, model training to application working are explained in section 4. Model evaluation is described in section 5. We concluded our work in section 6

Literature Survey

Table -1: Literature Survey

Paper	Writer Name	Description
1-PakhiChini: Automatic Bird Species Identification Using Deep Learning(IIEEE)(2020)	Kazi Md Ragib, Raisa Taraman Shithi, Shihab Ali Haq, Md Hasan, Kazi Mohammed Sakib, Tanjila Farah	In this paper propose a deep learning model that is capable of identifying individual birds from an input image.. Authors tend to additionally leverage pretrained ResNet models as pre-trained CNN networks with base models to encode the images. We achieved a top-5 accuracy of 97.98% on our classifications
2- Bird Species Identification using Deep Learning on GPU Platform(IEEE)(2020)	Pralhad Gavali,J.Saira Banu	In this experiment for training purpose 500 labeled and 200 unlabeled data are used for testing. For classification, Deep Convolutional Neural Networks was used. Final results show that the DCNN algorithm can be predicted at 88.33% of bird species.
3-Learning Semantically Enhanced Feature for (FGIC) Fine-Grained Image Classification (IEEE)(2019)	Wei Luo, Hengmin Zhang, Jun Li, and Xiu-Shen WeiWei	In this paper the author achieves the sub-feature semantic by arranging feature channels of a CNN into different groups through channel permutation. Experiments verified the effectiveness of our approach and validated its comparable performance to the state-of-the-art methods with accuracy of 94%.
4- MobileNet Model for Classifying Local Birds of Bangladesh from Image Content Using Convolutional Neural Network(IEEE)(2019)	Md. Romyull Islam, Nishat Tasnim, Shaon Bhatta Shuvo	In this experiment MobileNet Model is used on a dataset of 5 species with size 100 images and at the end they get 100% accuracy in training and testing both.

5. EVALUATION

For the evaluation purpose, we have considered two parameters, which are accuracy and loss to assess model performance on a train, validate and test dataset. Where accuracy represents how accurately a model was able to predict the correct class label whereas Cross-entropy Loss Function shows how far our calculated output is from the actual output. From the Cross-entropy value, we can conclude how confident our model is in its output[11].

Despite having high accuracy, our model is facing high cross-entropy loss especially in the case of a validation dataset from which we can conclude that our model can correctly identify the class label but has low confidence in its prediction. It may be possible there are some images in the validation dataset that are accurately classified but the model has low confidence in its prediction as a result mean validation loss is high as compared to the train and test dataset loss[11].

Table 2. Evaluation Result

Dataset	Accuracy	Cross-entropy loss
Train	97.44%	0.107
Validate	94.30%	0.4939
Test	98.40%	0.0563

3. CONCLUSIONS

This study has used the InceptionV3[7] model for revealing an application that has data of 200 species and uses deep learning for data training and image processing. Images can be uploaded by the end-user from the system or by the camera as an input to perceive detailed information of birds. The current model has achieved an accuracy of 97.44% on training data and 98.40 % on testing data. This application can be adopted by tourists and regional people who want the information of bird's species present in the image.

In the future, we propose to work on more data, to classify the gender of species, create a mobile application, and Choropleth Map will present the bird's habitation more precisely. A Choropleth map also for presenting the migration routes that birds used during migration.

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