

Literature Survey

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Project Name	Digital Naturalist-AI Enabled tool for Biodiversity Researchers

Abstract :

The increasing availability of digital images, coupled with sophisticated artificial intelligence (AI) techniques for image classification, presents an exciting opportunity for biodiversity researchers to create new datasets of species observations. We investigated whether an AI plant species classifier could extract previously unexploited biodiversity data from social media photos (Flickr). We found over 60,000 geolocated images tagged with the keyword “flower” across an urban and rural location in the UK and classified these using AI, reviewing these identifications and assessing the representativeness of images. Images were predominantly biodiversity focused, showing single species. Non-native garden plants dominated, particularly in the urban setting. The AI classifier performed best when photos were focused on single native species in wild situations but also performed well at higher taxonomic levels (genus and family), even when images substantially deviated from this. We present a checklist of questions that should be considered when undertaking a similar analysis.

S.No	Title	Author	Technique Used	Advantage
1.	AI Naturalists Might Hold the Key to Unlocking Biodiversity Data in Social Media Imagery	Tom A. August, Oliver L. Pescott	Deep Learning And Convolutional Neural Networks (Cnns)	<ul style="list-style-type: none">❖ Images Are Spatially Clustered❖ The Urban Area Has a Lower Proportion of Well-Classified Images❖ The Rural Area Has a Higher Proportion of

				Images of Naturally Occurring, Native Plants
2.	Avoid Oversimplifications in Machine Learning: Going beyond the Class-Prediction Accuracy	Sung Yang Ho, Limsoon Wong, Wilson Wen Bin Goh	Data science, Class-prediction accuracy	Class-prediction accuracy provides a quick but superficial way of determining classifier performance.
3.	Rethinking the Inception Architecture for Computer Vision	Christian Szegedy, Vincent Vanhoucke, Sergey Ioffe	VGGNet, GoogLeNet.	<ul style="list-style-type: none"> ❖ Spatial Factorization into Asymmetric Convolutions ❖ Factorization into smaller convolutions ❖ Utility of Auxiliary Classifiers ❖ Efficient Grid Size Reduction ❖ Model Regularization via Label Smoothing
4.	Machine learning to classify animal species in camera trap images: Applications in ecology	Michael A. Tabak, Mohammad S. Norouzzadeh, David W. Wolfson	Artificial intelligence, deep neural networks	<ul style="list-style-type: none"> ❖ Deep learning models that automatically classify wildlife ❖ This model achieved the highest

				<p>accuracy (97.6%) to date in using machine learning to classify wildlife in camera trap images</p> <ul style="list-style-type: none"> ❖ The model can also be valuable for researchers studying other species by removing images without any animals from the dataset before beginning animal classification
5.	Machine learning for image based species identification	Jana Wäldchen, Patrick Mäder	Automated species identification, Computer vision, Convolutional neural network,	<ul style="list-style-type: none"> ❖ It is fast development and ubiquity of relevant information. ❖ Species identification tasks are manifold and were comprehensive. ❖ It is widely adopted

				approach for object detection.
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