



Animal Detection Using Yolo Coco Model

Mr.A.Jai Shankar¹, Mr.R.Sathish Kumar²

¹Student, MCA, Krishnasamy College Of Engineering And Technology , Cuddalore, India. E-mail:sjai20039@gmail.com

²,AssistantProfessor, Krishnasamy College Of Engineering And Technology, Cuddalore, India. E-mail:sathishkumar635@gmail.com

ABSTRACT:

Over the last few years, there has been a steady rise in number of reported human–animal conflicts. While there are several reasons for increase in such conflicts, foremost among them is the reduction in forest cover. Animals stray close to human settlements in search of food, and often end up raiding crops or preying on cattle. There are at times human casualties as well. Proficient, reliable, and autonomous monitoring of human settlements bordering forest areas can help reduce such animal–human conflicts. A broad range of techniques in computer vision and deep-learning has shown enormous potential to solve such problems. In this paper, a novel, efficient, and reliable system is presented which automatically detects wild-animals using computer vision. The proposed method uses the YOLO object detection model to ascertain presence of wild animals in images. The model is fine-tuned for identifying ten different types of animals (Dog, Horse, Butterfly, elephant, Hen, Cat, Cow, Sheep, Spider, and Squirrels). The proposed method achieves an accuracy of 98.8% and 99.8% to detect animals.

Introduction:

In the recent few years, diverse research work happened to develop a practical approach to accelerate the development of deep learning methods. Numerous developments accomplished excellent results and followed by continuous reformations in deep learning procedures. Object localization is the identification of all the visuals in a photograph, incorporating the precise location of those visuals. By using deep learning techniques for object identification and localization, computer vision has reached a new zenith. Due to significant inconsistencies in viewpoints, postures, dimensions, and lighting positions, it is challenging to succeed in the identification of objects perfectly. Accordingly, considerable concern has been given by researchers to this area in the past few years. These techniques create region proposal networks (RPN), and then the region proposals are divided into categories afterward. On the other side, object detection algorithms using regression includes SSD and YOLO, etc. These methods also generate region proposal networks (RPN) but divide these region proposals into categories at the moment of generation. All of the procedures mentioned above have significant accomplishments in object localization and recognition. YOLO consolidates labels in diverse datasets to form a tree-like arrangement, but the merged labels are not reciprocally exclusive. research in the field. The main intent of this paper is to describe the design for a computer vision system, capable of detecting animals. DNN's could be leveraged to detect the presence of animals in the captured images. In addition to detecting the presence of an animal, in order to effectively track them and monitor their actions, it is also necessary to localize the animals within the image. This is the task of object detection. Object detection systems predict regions of interest within images, and in addition classify entities within these regions. Thus, object detection is the ideal choice for the system proposed in this paper. This paper introduces a novel method of reducing human animal conflicts, through constant and automatic monitoring of vulnerable areas using a system of cameras. Artificial intelligence (AI) is the ability of a computer program or machine to think and learn. It is also a field of study which tries to make computers "smart". As machines become increasingly capable, mental facilities once thought to require intelligence are removed from the definition. AI is an area of computer sciences that emphasizes the creation of intelligent machines that work and react like humans. Some of the activities computers with artificial intelligence are designed for include: Face recognition, Learning, Planning, Decision making etc. Artificial intelligence is the use of computer science programming to imitate human thought and action by analysing data and surroundings, solving or anticipating problems and learning or self-teaching to adapt to a variety of tasks. Object Detection and Tracking. There is a wide range of computer vision tasks benefiting society such as object classification, detection, tracking, counting, Semantic Segmentation, Captioning image, etc. Process of identifying objects in an image and finding its position is known as object detection. Various object detection tasks. With advancements in field of computer vision assisted by AI, realization of tasks was realizable along t time scale. Semantic segmentation task of clustering pixels based on similarities. Classification + Localization and object detection method of identifying class of object and drawing a bounding box around it to make it distinct. Instance segmentation is semantic segmentation applied to multi objects. The general intuition to perform the task is to apply DNN over the image. DNN works on image patches to carry out the task many such salient regions can be obtained by Region-Proposal Networks like Region Convolution Neural network (RCNN), Fast-Region Convolutional Neural Network (Fast-RCNN), Faster- Region Convolutional Neural Network (Faster-RCNN). To perform selective search for object recognition Hierarchical Grouping Algorithm is used. Few bottlenecks by these approaches are mitigated by state-of-the-art algorithms like You Only Look Once (YOLO), Single shot Detector (SSD). The efficient object detection algorithm is one which assures to give bounding box to all objects of vivid size to be recognized, with great computational capabilities, faster processing. YOLO and SSD assure to render promising results, but have a tradeoff between speed and accuracy. Hence, selection of algorithm is application specific.

EXISTING SYSTEM

Our Existing system describes a system for segmentation of animals from images.

- ▶ The procedure employed uses a multi-level iterative graph cut to generate object region proposals and accurately recognize regions of interest.
- ▶ This is especially useful when the animal blends together with the background and is difficult to identify.
- ▶ These proposals segmented into background and foreground in the second stage. Feature vectors are extracted from each image using AlexNet.

DISADVANTAGES OF THE EXISTING METHOD

- i. It cannot give an accuracy value.
- ii. Less storage
- iii. Less performance

PROPOSED SYSTEM

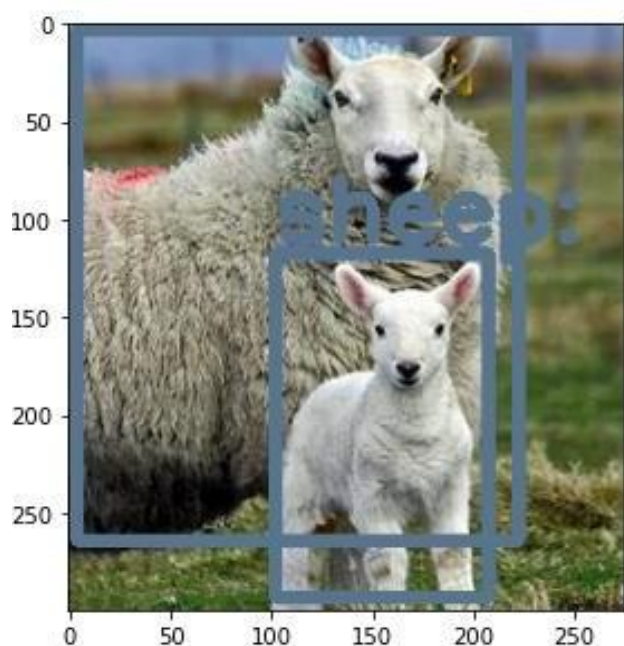
- The proposed method uses the YOLO object detection model to ascertain presence of wild animals in images.
- The model is fine-tuned for identifying Ten different types of animals (Dog, Horse, Butterfly, elephant, Hen, Cat, Cow, Sheep, Spider, Squirrels.)
- The proposed method achieves an accuracy of 98.8% and 99.8% to detect animals respectively.

The system proposed in this paper uses a network of cameras, connected to PIR motion sensors, so that image capture is triggered only when some movement is detected. This enables power conservation. The images captured through these cameras are processed to detect presence of wild animals, and if an animal is found, identify the species. Once identified, the animals are tracked for a suitable time in order to determine their intent – such as to find whether they are moving across the village, or into it. In the latter case, alerts are generated and local authorities are notified through proper channels. Understanding the intent goes a long way to reduce false positives, either due to a false detection or when there is no actual threat posed due to presence of the animal.

ADVANTAGES

- Easily detection with the help of Yolov3 and SSD
- High Accuracy
- Good Performance

RESULT:



FUTURE ENHANCEMENT

Continue to concentrate on expressive search including fuzzy keyword search, semantic keyword search, and so on. Besides, the secure channel utilized in our LFGS system should be eliminated as secure channel will incur high communication burden. Still need to further improve the efficiency of LFGS system so that it can be applied in various schemes.

CONCLUSION:

The proposed system detects the animals using computer vision. The proposed system is cost-effective and highly efficient, with an average accuracy of 98.8% in detecting and identifying animals in images. Although the prototype described in this paper is trained to recognize ten different species of animals, it is easily extendable to detect other types of animals with sufficient training data. The choice of species can also be region-specific, thereby providing a unique edge over other existing solutions. Such a system if implemented on a large scale, has potential to detect large data of animal categories.

REFERENCES

1. Caja, Gerardo, J. J. Ghirardi, M. Hernández-Jover, and D. Garín. "Diversity of animal identification techniques: From 'fire age' to 'electronic age'." *ICAR Technical Series* 9 (2004): 21-39.
2. Donovan, John, and Patricia Brown. "Animal identification." *Current protocols in immunology* 73, no. 1 (2018): 1-5.
3. Eradus, Wim J., and Mans B. Jansen. "Animal identification and monitoring." *Computers and Electronics in Agriculture* 24, no. 1-2 (2019): 91-98.
4. Voulodimos, Athanasios S., Charalampos Z. Patrikakis, Alexander B. Sideridis, Vasileios A. Ntafis, and Eftychia M. Xylouri. "A complete farm management system based on animal identification using RFID technology." *Computers and electronics in agriculture* 70, no. 2 (2020): 380-388.
5. Kellenberger, B., Volpi, M., Tuia, D.: Fast animal detection in uav images using convolutional neural networks. In: 2017 IEEE International Geoscience and Remote Sensing Symposium (IGARSS), pp. 866–869 (2017)
6. Krizhevsky, A., Sutskever, I., Hinton, G.E.: ImageNet classification with deep convolutional neural networks. In: F. Pereira, C.J.C. Burges, L. Bottou, K.Q. Weinberger (eds.) *Advances in Neural Information Processing Systems* 25, pp. 1097–1105. Curran Associates, Inc. (2012)
7. Liu, W., Anguelov, D., Erhan, D., Szegedy, C., Reed, S., Fu, C.Y., Berg, A.C.: Ssd: Single shot multibox detector. *Lecture Notes in Computer Science* pp. 21–37 (2016)
7. Lowe, D.G.: Distinctive image features from scale-invariant keypoints. *International Journal of Computer Vision* 60(2), 91–110 (2004)
8. Lukešič, A., Vojtíšek, T., Čehovin Žajc, L., Matas, J., Kristan, M.: Discriminative correlation filter tracker with channel and spatial reliability. *International Journal of Computer Vision* 126(7), 671–688 (2018)
9. Matuska, S., Hudec, R., Benco, M., Kamencay, P., Zachariasova, M.: A novel system for automatic detection and classification of animal. In: 2014 ELEKTRO, pp. 76–80 (2014)
10. Norouzzadeh, M.S., Nguyen, A., Kosmala, M., Swanson, A., Palmer, M.S., Packer, C., Clune, J.: Automatically identifying, counting, and describing wild animals in camera-trap images with deep learning. *Proceedings of the National Academy of Sciences* 115(25), E5716–E5725 (2018)
12. Parham, J., Stewart, C., Crall, J., Rubenstein, D., Holmberg, J., Berger-Wolf, T.: An animal detection pipeline for identification. In: 2018 IEEE Winter Conference on Applications of Computer Vision (WACV), pp. 1075–1083 (2018)
13. Redmon, J., Divvala, S., Girshick, R., Farhadi, A.: You only look once: Unified, real-time object detection. In: 2016 IEEE Conference on Computer Vision and Pattern Recognition (CVPR), pp. 779–788 (2016).