



SMART OBJECT DETECTION SYSTEM ON WEBCAM USING SCALED YOLOv4

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Abstract: Object detection is the capacity required with the aid of using maximum laptop imaginative and prescient systems. The ultra-modern studies have been making terrific improvements in lots of areas. Object detection and monitoring have quite a few makes use of the detection method we're the use of is primarily based totally on YOLO. In this paper, we additionally speak of modern and potential packages of item detection in numerous fields. The consequences represented right here advise that this structure may be similarly evolved and utilized in face detection, face recognition, anomaly detection, crowd counting, safety surveillance, etc. The YOLOv4 item detection neural community placed at the CSP techniques cutes each up and down and applies to small and huge networks whilst preserving the most advantageous pace and accuracy. We recommend a community scaling technique that modifies no longer additionally the depth, width, decision but additionally the shape of the community.

Index Terms – Object Detection, YOLO, Anomaly detection, Neural Network

I. INTRODUCTION

Object detection is a generation related to laptop imaginative and prescient and photo processing. This area acknowledges and identifies times of precise gadgets of a delegated elegance in virtual photographs and videos. Object detection does the paintings of mingling those responsibilities and localizes and classifies one or greater gadgets all through a picture. When a consumer or expert mentions "item recognition", they often mean "item detection" Object detection tries to detect all occurrences of gadgets from recognized elegance such as people, cars or faces, all through photography or video. In general, only some examples are the gift of the idea. Still, there is a considerable range of feasible locations and scales at which they will occur that should become something to explore. Each photo detection is said with the call of the item it is being detected, and that is frequently as easy because of the item's position, location, and scale, or they are placed out of the call described in phrases of a rectangular container. In rare cases, position records are more exhaustive and include the center of a linear or non-linear transformation. For example, a face detector through face detection can also additionally calculate the eye, nose and mouth locations and the bounding container of the face. Deep learning is the current reform field of machine learning. Use structured machine learning methods that are computationally time- and cost-effective, supervised, and accurate. This paves the way to solve various complex problems in computing domain. It has achieved measurable performance in various applications and has made great progress. It is used to discover complex structures in infinite amounts of data using backpropagation algorithm.

II. RELATED WORKS

[1]Chengji Liu et al.'s (2019), confirmed that the version educated with the same old units does now no longer have excellent generalization capacity for the degraded photos and has terrible robustness. Then the version became educated on the use of degraded photos which led to progressed common precision. It became proved that the common precision for degraded photos became higher with inside the popular degenerative version in comparison to the same old version.

[2] Joseph Redmon et al (2019), In assessment with Object detection strategies that got here earlier than YOLO, like R-CNN, YOLO added an unmarried unified structure for regression pass photograph into bounding packing containers and locating magnificence chances for every box. This intended that YOLO finished a whole lot quicker and additionally supplied greater accuracy. It can also expect artwork correctly.

[3] Wenbo Lan et.al(2019), The YOLO v2 and YOLO-R community fashions had been examined at the check set of the INRIA statistics set. The experimental consequences show that the YOLO-R community version is advanced to the distinctive YOLO v2 community version. the range of detection frames reached twenty-five frames/s, primarily collection the need of period of time performance.

[4] Rumin Zhang et.al(2019), The photographs of the not unusual place limitations had been classified and used for schooling YOLOv4. The item cleared out is implemented to the worried impediment. Different kinds of scenes, which include pedestrians, chairs, cars, and so on, are tested to show the effectiveness of this impediment detection algorithm.

[5] Zhimin Mo et al.(2019),The YOLO set of rules proposed identifies the placement of the solder joints efficiently in real-time. This is useful to grow the performance of the manufacturing line and it has incredible for the power and real-time of the welding of car door panels.

[6] T. Ahmad et al.(2018),proposed Multi-Directional Car Plate Detection, using You Only Look Once (YOLO) architecture. The authors performed some refinements on the YOLO object detection architecture to make it work in situations that cameras have different degrees of rotation while taking videos or pictures. They call the model mentioned above, Multi-Directional YOLO (MDYOLO). The authors applied the attention method to estimate the precise location of the license plates. Afterward, the estimated region is passed to the MD-YOLO. The authors achieved acceptable results on different subsets of the AOLP1 dataset. They achieved the precision of 99.51 %, 99.43 %, and 99.46 % on Access Control, Traffic Law Enforcement, and Road Patrol subsets of AOLP, respectively.

[7] S. M. Silva et al.(2017), refined the FAST-YOLO network to extract both frontal views of the car and the license plate. They adjusted the Fast-YOLO to output both the car and license plate for the detection phase. For character detection and recognition, the authors modified YOLO architecture and applied a heuristic approach to improving their final result. The authors stated that they achieved 63.18 % accuracy, while the Sight hound achieved 55.47 % for correctly detected and recognized license plates.

[8] S. Venkateswarlu added ideas for optical character recognition (OCR) and text-to-speech (TTS) synthesizer for Raspberry pi. This gadget enables visually impaired people to effectively interact with computer systems through voice interfaces. Extracting text from color photos with laptop vision is a difficult task. The text-to-speech approach uses an OCR approach to scan and read letters and numbers in photos and convert them into speech. This tool includes a module, a photo processing module and a sound processing module. It has become a tool that is basically completely based on the Raspberry Pi v2 with a processor speed of 900 MHz.

III. EXISTING SYSTEM

Computer imaginative and prescient is the sphere of technology seeking to update human eyesight, spotting items with inside the actual global from virtual pictures or films the usage of enlightened techniques and reworking them into different representations. In the market, there are numerous systems, on the way to skip through those criteria. Many of them are part of the end result of a couple of studies projects, others are a unmarried manufactured from the economic sphere and they're now no longer consumer friendly. The goal is to create consumer friendly. The problems of the existing system are two priorities: object classification and object localization, real-time detection speed, multidimensional scales and aspect ratio, limited data, class imbalance.

IV. PROPOSED SYSTEM

In the proposed system, the Object detection system module performs the operations such as, Pre-processing of the image. Detecting objects from the given image. Display image/live with the placeholders set. Display detected objects. In the proposed system, Object tracking system module performs the operations such as, Find objects, Display live with placeholders set, Display detected objects, Keep track of objects. When compared to the existing system proposed system is much faster and accurate. YOLOv4- huge version achieves latest results: 56.5% AP (74.% AP50) for the MS COCO dataset at a pace of ~17 FPS on Tesla V100, whilst with the check time augmentation, YOLOv4-huge achieves 56.0% AP (73. three AP50). To the coolest of our knowledge, that is presently the very best accuracy at the COCO dataset amongst any posted work. The YOLOv4-tiny version achieves 22.0% AP (42.0% AP50) at a pace of ~443 FPS on GTX 2080 Ti, whilst through the usage of Tensor RT, batch size = four and FP16-precision the YOLOv4-tiny achieves 1774 FPS.

The figure 1 describes the architecture flow of the process. At the first stage, input images are captured and compared with the trained data set. Then it recognizes the object If the object is approved, it will display in the bounding box, it continues the process till the webcam has turned off. The object recognition model of the proposed system can track multiple people in real time moving in a given scene or frame. Real-time vehicle recognition models are critical to the success of autonomous vehicle systems. These systems must detect, locate and track objects around them in order to navigate safely and skillfully around the world. Standard ML-based methods use computer vision techniques to examine various image features, such as color histograms and edges, to identify groups of pixels that may belong to a particular object. These features are fed into a regression model to predict object locations along with labels.

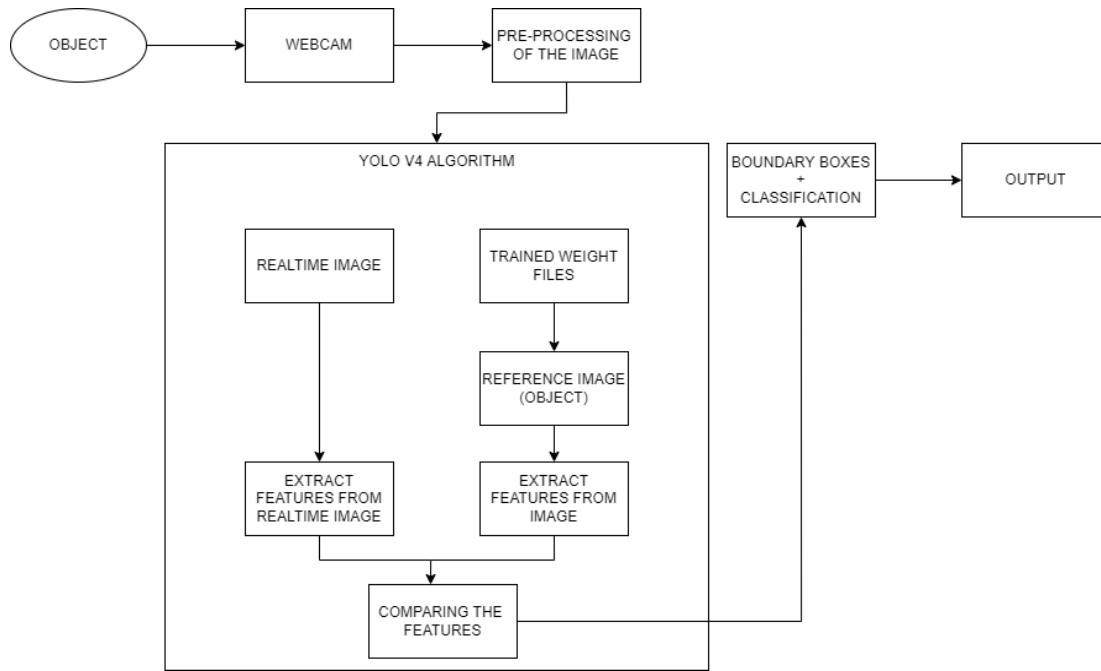


Figure 1: Architecture Diagram

V. RESULT

YOLOv4 Example on Test Image

Let's make sure our model has successfully been loaded and that we can make detections properly on a test image.

```
# run test on person.jpg image that comes with repository
image = cv2.imread("data/person.jpg")
detections, width_ratio, height_ratio = darknet_helper(image, width, height)

for label, confidence, bbox in detections:
    left, top, right, bottom = bbox2points(bbox)
    left, top, right, bottom = int(left * width_ratio), int(top * height_ratio), int(right * width_ratio), int(bottom * height_ratio)
    cv2.rectangle(image, (left, top), (right, bottom), class_colors[label], 2)
    cv2.putText(image, "{} [{}:{}]".format(label, float(confidence), (left, top - 5), cv2.FONT_HERSHEY_SIMPLEX, 0.5, class_colors[label], 2)
cv2.imshow(image)
```

Figure 2: Detection in test image

In the Figure 2, we are testing with the sample image which we have provided. It gives better result while testing with an image.

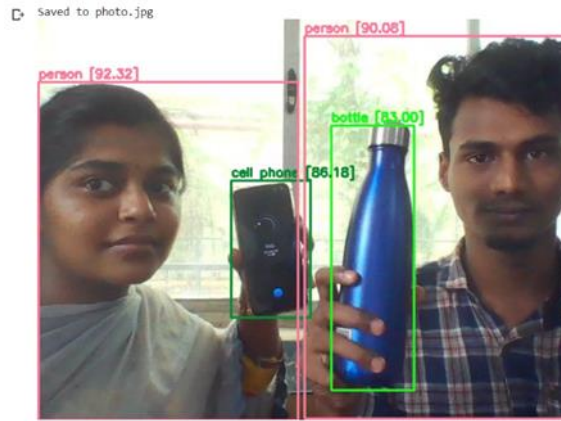


Figure 3: Detection on a webcam

In the Figure 3 we have tested on the laptop webcam it gives better result and accuracy and the detection of object is also very much faster and accurate. We have also tested by uploading the videos it gives more accurate results. We have done so for about 40 objects and it detects objects correctly which are in the frames.

VI. CONCLUSION

The planned object detection and chase algorithmic program were generally tested to control in an exceedingly complicated and planet. It was found to possess exceptional accuracy and preciseness of 97%. We've tested the proposed algorithm to trace assort objects against an atmosphere include littered things of assorted sizes, shapes, and colors. The implementation of the algorithm was found to be ultimately quick and strong and so this algorithm is best than that are accessible therefore far. The version may be used to be expecting body via way of means of body as they're captured via way of means of the digital camera till the loop receives interruption.

VII. FUTURE SCOPE

In the future, we are going to build a security system using this model. We are going to build a model for more than 80 objects that we use in our day today life. With this system, we can develop the driver less cars system for that we are going to detect more objects related to this work which can be extended by adding more constraints like alerting the driver by switching on the parking lot or slowing down the vehicle. This system is developed from a product point of view that will be highly helpful in avoiding accidents and saving lives.

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