



International Journal for Innovative Engineering and Management Research

A Peer Reviewed Open Access International Journal

www.ijiemr.org

COPY RIGHT



ELSEVIER
SSRN

2022 IJIEMR. Personal use of this material is permitted. Permission from IJIEMR must be obtained for all other uses, in any current or future media, including reprinting/republishing this material for advertising or promotional purposes, creating new collective works, for resale or redistribution to servers or lists, or reuse of any copyrighted component of this work in other works. No Reprint should be done to this paper, all copy right is authenticated to Paper Authors

IJIEMR Transactions, online available on 25th Jun 2022. Link

[:http://www.ijiemr.org/downloads.php?vol=Volume-11&issue= Spl Issue 05](http://www.ijiemr.org/downloads.php?vol=Volume-11&issue= Spl Issue 05)

DOI: 10.48047/IJIEMR/V11/SPL ISSUE 05/14

Title **WEAPON DETECTION USING MACHINE LEARNING**

Volume 11, SPL ISSUE 05, Pages: 89-98

Paper Authors

**Mr. P. Sudhakar , M. Ram Sri, Ch. Bhanu Sri, Y. Durga Ramanjaneya Kumar Babu ,
J. Yaswanth Devi Sagar**



USE THIS BARCODE TO ACCESS YOUR ONLINE PAPER

To Secure Your Paper As Per **UGC Guidelines** We Are Providing A Electronic Bar Code

WEAPON DETECTION USING MACHINE LEARNING

Mr. P. Sudhakar¹, M. Ram Sri², Ch. Bhanu Sri³, Y. Durga Ramanjaneya Kumar Babu⁴,
J. Yaswanth Devi Sagar⁵

¹Assistant Professor, Dept of CSE, ²18ME1A0559, ³18ME1A0513,
⁴18ME1A0524, ⁵18ME1A0541

Ramachandra College of Engineering, A.P..., India.

ramsrimutyala517@gmail.com , bhanu.challagolla@gmail.com,

durgaramanjaneya218@gmail.com

Abstract

Automatic control systems are becoming more important for security personnel as the number of criminal actions increases. A new model is developed in this work that uses deep learning to detect seven different weapon kinds. Based on the VGG Net planning, this model proposes a new approach to weapon classification. Assault rifles, bazookas, grenades, hunting rifles, knives, pistols, and revolvers are all taught to the model. The suggested model is developed using the Keras package and the TensorFlow platform. To establish the training success rate, select the training technique, produce layers, complete the training process, preserve training in the computer environment, and test the trained model, a new model is utilized. To prepare the model network for training, a new dataset is created that includes seven different weapon kinds. The suggested model is compared to the VGG-16, ResNet-50, and ResNet-101 models using this dataset to see which one produces the best classification results. The projected model's success accuracy of 98.40 percent is higher than the VGG-16 model's 89.75 percent success accuracy, the ResNet-50 model's 93.70 percent achievement accuracy, and the ResNet-101 model's 83.33 percent achievement accuracy as a result of the comparison.

1 Introduction

People employ surveillance cameras in various regions to prevent crime in the age of modern science and technology. Several video systems are installed in various regions, and security personnel must simultaneously monitor all of these cameras. Security guards typically come to the scene after a crime has occurred, review the recorded photos, evaluate the photographs, and gather the necessary evidence, consequently it is necessary to develop a proactive system at the crime scene. If the software warns security guards as soon as hazardous things are detected, appropriate action can be done to prevent the potential criminal from binding a crime. As a result, it's critical to develop a system that can discover ways to locate threatening gadgets.

The function of deep gaining knowledge in enhancing undertaking overall show in safety

controls systems is considered undeniable. Deep learning is a sub-field of system learning. It makes use of many layers of non-linear dispensation devices for deep gaining knowledge of characteristic removal and conversion [10]. The deep studying structure is founded on the studying of a couple of feature degrees of information. Deep studying is based on getting to know from the illustration of the fundamental records. The illustration of a photo can be measured to comprise a vector of density values in line with pixels or capabilities such as clusters of edges and custom shapes, with some representing the records higher. The primary architecture of the deep gaining knowledge of concept is the convolutional neural community (CNN), which consists of difficulty, pooling, activation characteristic, dropout, absolutely related, and category layers. Within the last few years, deep studying has ended up a strength within the discipline of item finding

and classification and Picture division. To this point, CNN's have completed the high-quality consequences for classical image processing troubles, together with photo segmentation, class, and detection. Nowadays, maximum criminal sports have finished the usage of handheld guns. Many studies have revealed that hand-held weapons are the maximum critical crook factors used for diverse crimes, consisting of theft, unlawful looking, and terrorism. Installing a surveillance device or control cameras to allow protection units to take necessary measures at early tiers is one way to reduce such crook activities. Weapon identification is difficult due to the many intricacies involved. Self-occlusion and similarities between gadgets and backdrop structures are the most crucial challenges in weapon detection.

When a section of the gun is blocked on one side, this is known as self-occlusion. Similarity among gadgets takes place while one-of-a-kind gadgets along with hands and garments look like weapons. Heritage issues seek advice from the ones related to the heritage against which the gun is placed.

This examination proposes a brand-new version to come across seven different weapon kinds (assault rifles, bazookas, grenades, looking rifles, knives, pistols, and revolvers) based on the deep gaining knowledge of technique. Gun violence has a tremendous impact on the public's health, psychological well-being, and economic well-being.

Every year, a large number of people are killed by gun violence. Children who are bare to high levels of violence in their societies or from side to side the media are more likely to experience psychological trauma. Children who are uncovered to gun-related violence, whether as victims, perpetrators, or bystanders, can suffer short- and long-term psychological consequences. According to numerous studies, the portable gun is the major weapon used in a variety of crimes such as break-ins, robberies, shoplifting, and rape. these crimes can be decreased by detecting disruptive conduct

early on and thoroughly checking suspicious actions so that law enforcement officials can respond quickly [1].

Although the human visual system is quick and precise, and can also perform complex tasks like distinguishing different items and recognising snags with little conscious thought, it is a well-known fact that if an individual watches something very similar for an extended period of time, there is a risk of sluggishness and lack of regard.

2 Related Works

A study investigating an automatic pistol detection device in films for surveillance and protection purposes [18] created basic training statistics using the consequences of the deep CNN category. The high-quality organization model changed explored to minimize fake positives utilizing comparing category models based totally on the sliding window and area thought approaches, and it was seen that the best effects had been acquired from the quicker region-based CNN model. In a photograph-fusion-based take a look, inequity map shrewdness and assortment of candidate regions from input frames had been finished. The weapon detection in surveillance machines the usage of the yolov3 algorithm is faster than the previous CNN, R-CNN, and quicker CNN algorithms. Popular this period where belongings are automated, item detection will become one of the most exciting fields. While it involves item detection in surveillance structures, speed plays a critical role in locating an item fast and alerting the expert. This effort tried to reap the similar and its capable of producing a quicker result compared to the previously current structures. The longer-time period effort of the future device is to extend a more quantity of types of weapons and classify them. The correctness of the weapon noticed can be advanced by the use of extraordinary forms of algorithms. A likely way to enhance this painting is to discover a hidden weapon that cannot be detected by the usage of the regular digicam. Also, analyzing the behavior of the humans to locate any suspicious sports like hiding the

weapon can be completed to improve this surveillance machine. The alert machine also can be stepped forward to inform a couple of users if the weapon is detected. A surveillance gadget with these features may be useful to prevent violent crimes and provide protection to the public.

A low-value, symmetrical dual camera machine changed into established to take advantage of these statistics. Consistent with the effects, the number of fake positives decreased and the detection of items in surveillance videos has become handier. Brightness-guided preprocessing, referred to as darkening and contrasting at gaining knowledge of and check stages, became proposed to enhance detection best in surveillance videos and a chilly steel weapon detection version became offered. In a hybrid weapon detection study the use of fuzzy common sense, a machine was evolved to hit upon dangerous objects including weapons and knives the use of additional parameters that progressed correct effects and decreased fake alarm rates.

The objective of this research is to create a smart surveillance security structure that can detect weapons, specifically guns. With the consistently expanding prevalence of distributed computing, the interest for distributed storage has additionally expanded dramatically. Processing firms are presently not the main customers of distributed storage and distributed computing, yet rather normal organizations, and even end-clients, are exploiting the colossal capacities that cloud administrations can give. While partaking in the adaptability and accommodation brought by distributed storage, cloud clients discharge command over their information, and especially are frequently incapable to find the real their information; The weapon detection in surveillance machines the usage of the yolov3 algorithm is faster than the previous CNN, R-

CNN, and quicker CNN algorithms. Popular this period where belongings are automated, item detection will become one of the most exciting fields. While it involves item detection in surveillance structures, speed plays a critical role in locating an item fast and alerting the expert. This effort tried to reap the similar and its capable of producing a quicker result compared to the previously current structures. The longer-time period effort of the future device is to extend a more quantity of types of weapons and classify them. The correctness of the weapon noticed can be advanced by the use of extraordinary forms of algorithms. A likely way to enhance this painting is to discover a hidden weapon that cannot be detected by the usage of the regular digicam. Also, analyzing the behavior of the humans to locate any suspicious sports like hiding the weapon can be completed to improve this surveillance machine. The alert machine also can be stepped forward to inform a couple of users if the weapon is detected. A surveillance gadget with these features may be useful to prevent violent crimes and provide protection to the public.

For this, we used a combination of compute vision and deep learning to identify a weapon from a recorded image.

Recent advances in the fields of machine learning and deep learning, particularly convolutional neural networks, have shown significant improvement in the domains of picture object detection and recognition. Object detection and categorization are crucial for additional object tracking duties in any video surveillance application.

This project aims to build and construct a system that can quickly identify firearms, rifles, and fire while using little computer resources. Most human-assisted applications are now automated and computer-based, as evidenced by technology breakthroughs. These computer-based systems will eventually be superseded by smart machines, robots, or humanoid robots. Object detection is critical for comprehending and interpreting items in

order to offer robots with visionary sense. this, our proposed technology may be used to detect any weapon or dangerous assets in surveillance and security robots.

In this work, we aimed to design an integrated framework for reconnaissance security that gradually identifies the weapons, and if the identification is positively true, it will warn/brief the security personnel on how to handle the situation by arriving at the scene via IP cameras. We offer a concept that gives a machine a visionary sense to detect dangerous weapons and can also inform a human administrator when a pistol or firearm is visible near the edge. We've also programmed an entryway locking system for when the shooter appears to be wielding a potentially lethal weapon. If possible, we may also share the live image with security personnel so they can make the move in the meantime using IP webcams. The objective of this research is to create a smart surveillance security structure that can detect weapons, specifically guns. With the consistently expanding prevalence of distributed computing, the interest for distributed storage has additionally expanded dramatically. Processing firms are presently not the main customers of distributed storage and distributed computing, yet rather normal organizations, and even end-clients, are exploiting the colossal capacities that cloud administrations can give. While partaking in the adaptability and accommodation brought by distributed storage, cloud clients discharge command over their information, and especially are frequently incapable to find the real their information; The weapon detection in surveillance machines the usage of the yolov3 algorithm is faster than the previous CNN, R-CNN, and quicker CNN algorithms. Popular this period where belongings are automated, item detection will become one of the most exciting fields. While it involves item detection in surveillance structures, speed plays a critical role in locating an item fast and

alerting the expert. This effort tried to reap the similar and its capable of producing a quicker result compared to the previously current structures. The longer-time period effort of the future device is to extend a more quantity of types of weapons and classify them. The correctness of the weapon noticed can be advanced by the use of extraordinary forms of algorithms. A likely way to enhance this painting is to discover a hidden weapon that cannot be detected by the usage of the regular digicam. Also, analyzing the behavior of the humans to locate any suspicious sports like hiding the weapon can be completed to improve this surveillance machine. The alert machine also can be stepped forward to inform a couple of users if the weapon is detected. A surveillance gadget with these features may be useful to prevent violent crimes and provide protection to the public. The objective of this research is to create a smart surveillance security structure that can detect weapons, specifically guns. With the consistently expanding prevalence of distributed computing, the interest for distributed storage has additionally expanded dramatically. Processing firms are presently not the main customers of distributed storage and distributed computing, yet rather normal organizations, and even end-clients, are exploiting the colossal capacities that cloud administrations can give. While partaking in the adaptability and accommodation brought by distributed storage, cloud clients discharge command over their information, and especially are frequently incapable to find the real their information; The weapon detection in surveillance machines the usage of the yolov3 algorithm is faster than the previous CNN, R-CNN, and quicker CNN algorithms. Popular this period where belongings are automated, item detection will become one of the most exciting fields.

In a weapon class that examines advanced usage of a deep CNN, new tactics were

presented. Inside the proposed technique, the weights of the pre-educated VGG-sixteen version had been taken. Using this version, the effects of converting the variety of neurons inside the related layer on type had been investigated. In a study aiming to hit upon firearms in surveillance films, the point of interest became placed most effective in the best areas where human beings were observed [5] and a weapon detection device changed into applied using the separate additives of the guns. An observation on multilevel security management provided a gadget for the control of multimedia security in the internet of things structures. This gadget mechanically analyzed multimedia occasions and calculated protection degrees [28]. In any other real-time object detection look, the authors detected hand-held guns (pistols and rifles). In that take a look at, a TensorFlow-based software of overfeat, a CNN-based picture classifier, and feature. In the areas of object identification and recognition, recent work in the fields of deep learning and transfer learning has shown tremendous success.

3 Implementation

3.1 Pre-Processing and Dataset

Since there is no standard dataset for weapon discovery and acknowledgment, 5214 weapon pictures were ready by downloading them from the web. Downloaded weapon pictures should be of good quality pictures with various points to be effective in identifying and perceiving genuine weapons. Furthermore, to accomplish higher achievement exactness from the created brain network model, insignificant articles in every weapon pictures were taken out. In that capacity, downloaded weapon pictures were inspected exclusively and different PC application programs were utilized to make changes to the pictures, for example, cushioning, veiling, foundation cleaning, measuring, and pivot, agreeing to the substance. After getting ready separate pictures for every weapon class, they were changed over into a dataset. The weapon types and quantities of pictures having a place with the dataset utilized in the review are given.

The dataset included pictures of attack rifles, bazookas, projectiles, hunting rifles, blades, guns, and pistols (Figure 1). The pictures utilized in the dataset were changed into a dark variety design at 144×144 pixels utilizing the Python programming language. Then, at that point, the pictures having a place with the dataset were named under the weapon class to which they had a place and assembled.



Figure 1. Sample dataset of different classes: (a) assault rifles; (b) bazookas; (c) grenades; (d) hunting rifles; (e) knives; (f) pistols; (g) revolvers.

3.2 CNN Model

In object acknowledgment applications, the most utilized profound learning calculation is the CNN calculation. This calculation presented remarkable progress in the ImageNet Large-Scale Visual Recognition Competition thought in 2012 and has been utilized in the advancement of new models in numerous areas from that point forward. In the ongoing review, another model is proposed given the VGG-16 model (Figure 3) [21]. The model comprises of an aggregate of 25 layers (convolution ($n = 7$), pooling ($n = 4$), dropout ($n = 4$), redressed direct units (ReLU) ($n = 7$), level ($n = 1$), completely associated ($n = 1$), and arrangement ($n = 1$)) furthermore, 337,671 boundaries. The objective of this research is to create a smart surveillance security structure that can detect weapons, specifically guns. With the consistently expanding prevalence of distributed computing, the interest for distributed storage has additionally expanded dramatically. Processing firms are presently not the main customers of distributed storage and

distributed computing, yet rather normal organizations, and even end-clients, are exploiting the colossal capacities that cloud administrations can give. While partaking in the adaptability and accommodation brought by distributed storage, cloud clients discharge command over their information, and especially are frequently incapable to find the real their information; The weapon detection in surveillance machines the usage of the yolov3 algorithm is faster than the previous CNN, R-CNN, and quicker CNN algorithms. Popular this period where belongings are automated, item detection will become one of the most exciting fields.

item detection will become one of the most exciting fields. While it involves item detection in surveillance structures, speed plays a critical role in locating an item fast and alerting the expert. This effort tried to reap the similar and its capable of producing a quicker result compared to the previously current structures. The longer-time period effort of the future device is to extend a more quantity of types of weapons and classify them. The correctness of the weapon noticed can be advanced by the use of extraordinary forms of algorithms. A likely way to enhance this painting is to discover a hidden weapon that cannot be detected by the usage of the regular digicam. Also, analyzing the behavior of the humans to locate any suspicious sports like hiding the weapon can be completed to improve this surveillance machine. The alert machine also can be stepped forward to inform a couple of users if the weapon is detected. A surveillance gadget with these features may be useful to prevent violent crimes and provide protection to the public. The objective of this research is to create a smart surveillance security structure that can detect weapons, specifically guns. With the consistently expanding prevalence of distributed computing, the interest for distributed storage has additionally expanded dramatically. Processing firms are presently not the main customers of distributed storage and distributed computing, yet rather normal organizations, and even end-clients, are exploiting the colossal capacities that cloud administrations can give. While partaking in the adaptability and accommodation brought by distributed storage, cloud clients discharge command over their information, and especially are frequently incapable to find the real their information;

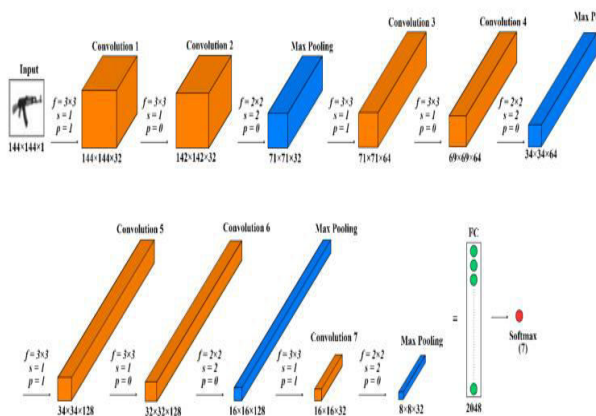


Figure 2. Developed convolution neural network model (f: filter size, s: stride, p: padding, FC: fully connect

The number of layers while planning a model organization and the boundary values utilized in these layers altogether influence the preparation season of the brain organization, the expense of handling, and the decision of gadget to be applied. In this review, the model we proposed was liked for the issue of weapon discovery and acknowledgment because of the modest number of layers contrasted with the exemplary VGGNet model, the capacity to prepare on minimal expense PCs, and the low preparation time, and the high achievement exactness. The weapon detection in surveillance machines the usage of the yolov3 algorithm is faster than the previous CNN, R-CNN, and quicker CNN algorithms. Popular this period where belongings are automated,

In the model introduced in Figure 2, two convolution layers and a most extreme pooling layer are applied to the information picture in a dark variety design. In the pooling layer, the two-venture 2×2 channel framework is moved over the information lattice to make another grid esteem. Greatest pooling holds

the new lattice esteem by taking the most extreme worth of each square to better gauge the picture. Furthermore, the ReLU actuation work is utilized in convolution layers. This is a non-straight capacity that replaces all regrettable pixel values with a zero worth, which guarantees that the organization is more proficient and quicker [34]. Utilizing a 25% dropout layer after each pooling layer, remembrance is forestalled [35]. Convolution, pooling, and neglecting processes are rehashed with various boundaries and different channel direct qualities in later layers. Then, at that point, neurons are shaped into an exhibit by straightening and completely associating layers. Because of the fixing system, 2048 neurons are acquired by shaping a completely associated layer. The last layer of the model is the arrangement layer with the softmax enactment work. In this layer, the resulting esteem is gotten by ordering the quantity still up in the air in the dataset. Because of the characterization, seven unique weapon types take worth in the scope of 0-1 and the weapon type with the most noteworthy worth presents the assessed esteem.

A CNN is a deep learning method that can take a raw input image and assign learnable weights and biases to the image's many aspects/objects. In a CNN model, a convolutional layer is in charge of extracting high-level characteristics such as edges from the input image. This works by repeatedly applying the kxk kernel filter on the raw image. this produces activation maps or feature maps as a consequence. The presence of detected features from the given input is represented by these feature maps.

Compared to other classification algorithms, the amount of preprocessing required is substantially lower with us; in the usual technique, filters are hand-engineered, but in CNN, these are learned through a series of iterations and training. reliably upholds an area mindful information stacking and stockpiling by relegating information hubs as indicated by user specified protection

approaches effectively tracks and progressively amends potential information relocation (because of adjusting or information replication needs) inside the group that could abuse information arrangement strategies recognizes possibly unlawful information movement, by observing attachment correspondence between individual information hubs and relating it with the limitations forced the strategy.

3.3 Media and Libraries Used

The model proposed in this study was created utilizing the Keras library in light of TensorFlow. Recorded as a hard copy of the codes, the NumPy, Matplotlib, PIL, Os, OpenCV, Sklearn, and what's more, Imageio libraries were additionally used. TensorFlow is a connection point for preparing and carrying out AI calculations. It is additionally an open-source programming library that performs mathematical computations utilizing information stream designs. Calculation utilizing TensorFlow is attempted in numerous frameworks going from cell phones to enormous scope conveyed frameworks, and it is likewise used to communicate different calculations created for CNN models. TensorFlow can be set up with two distinct processors for illustrations handling unit (GPU) or on the other hand focal handling unit (CPU) support. TensorFlow projects commonly run quicker on GPUs than on CPUs. In this review, the model was prepared on a PC furnished with an Intel Core i7-9750H 2.60 GHz processor with the 896 Compute Unified Device Architecture, Nvidia GeForce GTX 1650 illustrations card, and 8 GB RAM. The review meant to propose a model with an emphasis on giving the most noteworthy exactness, awareness, and particularity rates, while the most minimal misfortune rates were gotten utilizing CNN. Utilizing the new model, the strategy expected for still up in the air, model layers were made, the preparation cycle was applied and kept in the PC climate, the

achievement pace the not entirely set in stone, and the prepared model was tried. Moreover, the model was executed by composing codes in the Python programming language. TensorFlow is a connection point for preparing and carrying out AI calculations. It is additionally an open-source programming library that performs mathematical computations utilizing information stream designs. Calculation utilizing TensorFlow is attempted in numerous frameworks going from cell phones to enormous scope conveyed frameworks, and it is likewise used to communicate different calculations. While the most minimal misfortune rates were gotten utilizing CNN. Utilizing the new model, the strategy expected for still up in the air, model layers were made, the preparation cycle was applied and kept in the PC climate, the achievement pace the not entirely set in stone, and the prepared model was tried. The possibility of our methodology is that, whenever information is dispensed per clients' area inclinations, our system screens continuous record moves in the cloud and is fit for recognizing expected unlawful exchanges. An unlawful exchange in our setting signifies moving touchy information outside the lawful limits indicated by the document proprietor (e.g., putting away a record in an actual area other than whatever the document proprietor wants). Our methodology expands on the perception that clients' area inclinations are frequently predictable with security regulations and guidelines. Accordingly, records can be accumulated into bunches in which various clients share the comparable, on the off chance that not the equivalent, area inclinations. In like manner, our framework designates cloud hubs in view of the comparability of clients' area inclinations. All the more explicitly, we model the document moves among hubs as a weighted diagram and afterward

augment the likelihood that records with comparative protection inclinations will be put away in a similar district. We then devise attachment checking capacities to screen the ongoing correspondence among cloud hubs.

4 Results

In the review, tests were directed for seven unique weapon types and the suggested model was contrasted and the VGG-16, ResNet-50, and ResNet-101 models. Taking all things together in four models, the dataset that was utilized in the preparation of the organization was isolated into three gatherings, 60% as preparing, 20% as challenging, and 20% as the approval dataset. This present circumstance is displayed in Table 2. The preparation and testing datasets were utilized during the preparation of the organization. The achievement pace of the organization was acquired by testing the model with an approval dataset that it had never seen during preparation.

During the preparation processes in every one of the four models, the boundary upsides of the actuation work (ReLU) [34], small scale group size (32), dropout (0.25) [35], streamlining calculation (Adamax) [36], and ages (30) were utilized something similar. The information picture size for these four model instructional exercises was set to 144×144 pixels in dim variety design. In each of the four models, "precision", "misfortune", "awareness", and "explicitness" change charts decide the progress of the organization during preparing, which is displayed. Whenever the diagrams of the VGG-16 model given in Figure 4a were analyzed, it was seen that the model organization didn't pick up during the initial 6 ages and learned with a 90.12% achievement precision because of 30 ages. Comparable outcomes can likewise be seen in the misfortune, awareness, and particularity charts. Figure 4b shows that the ResNet-50 model started to get familiar with the model organization after the primary age and arrived at a triumph precision of 94.25% because of

30 ages. As displayed in Figure 4c, The ResNet-101 model came to a triumph precision of 84.43% because of 30 ages and had a lower achievement exactness than the ResNet-50 model. In Figure 4d, it very well may be seen that the model organization started to learn after the principal age for the proposed model and arrived at a triumph exactness of 98.32% because of 30 ages. Comparable outcomes can likewise be seen in the misfortune, responsiveness, and particularity diagrams for this prototypical. The progress of the model organization relies upon the layers, what's more, boundary values are utilized in the brain organization. In the suggested convolutional brain network model, handling the information and concentrating the essential features is significant. Aimed at these reasons, the coatings and boundary values utilized in the prototypical are chosen by the model organization, which expands the achievement rate. In the suggested model, the quantity of layers utilized was not exactly for different models, and in like manner the total boundaries was decreased; in any case, the proposed model can process, train, and test information quicker than other models. Accordingly, the proposed model seems to have higher achievement precision and a lower misfortune rate than the VGG-16, ResNet-50, and ResNet-101 models. What's more, with the future model, model organizations were tried utilizing 1043 weapon pictures in the organization approval bunch to dissect the proficiency of the VGG-16, ResNet-50, and ResNet-101 models, by the "exactness", "responsiveness", "particularity", then "misfortune" values for every one of the four models given in Table 3.

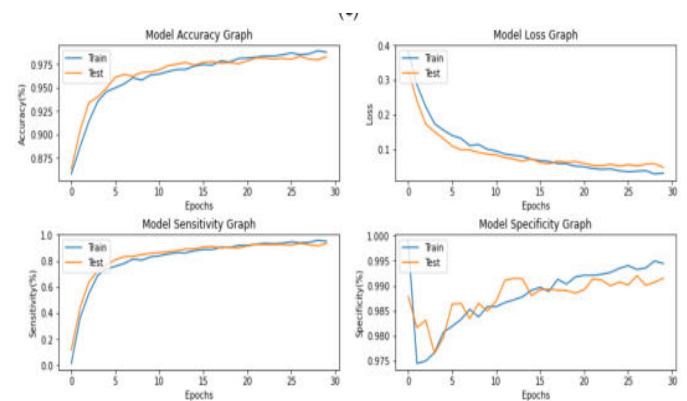
Whenever the diagrams assumed in Figure 4 and the qualities assumed in Table 3 are inspected, it should be visible that the future model advanced quicker beginning from the main ages and needed a advanced achievement degree. A disarray lattice was additionally made to break down the arrangement adequacy of the proposed perfect. As per this chart, the best characterization charges were acquired for

attack rifles at 99.45% and hunting rifles at 99.45%, while the arrangement charges were least fruitful for pistols at 94.62% and bazookas at 97.72%. Albeit the number of data points for the bazooka weapon type was not exactly accessible for the leftover weapon types in the dataset, it was even additional effectively characterized than the pistol type. This can remain ascribed to the visual likenesses among guns and guns from numerous perspectives.

The proposed model, as well as the VGG-16, ResNet-50, and ResNet-101 models, have high training success rates.

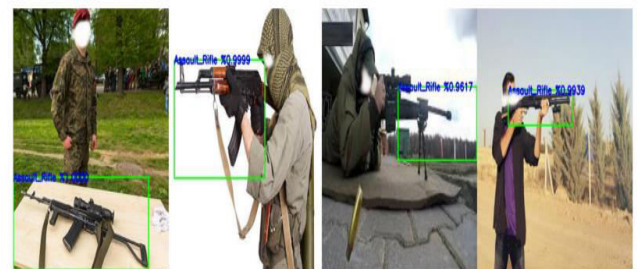
Measures	VGG-16 Values	ResNet-50 Values	ResNet-101 Values	Proposed Model Values
Accuracy	89.75%	93.70%	83.33%	98.40%
Sensitivity	89.71%	76.48%	41.05%	92.89%
Specificity	98.84%	96.55%	90.44%	99.28%
Loss	0.638	0.279	1.615	0.052

Graphs depicting changes over time: (a) the VGG-16 model; (b) the ResNet-50 model; (c) the ResNet-101 model; (d) the proposed model.



5 Output Samples

Sample test results for different classes: (a) assault rifles; (b) bazookas; (c) grenades; (d) hunting rifles; (e) knives; (f) pistols; (g) revolvers.



Conclusion

The weapon detection in surveillance machines the usage of the yolov3 algorithm is faster than the previous CNN, R-CNN, and quicker CNN algorithms. Popular this period where belongings are automated, item detection will become one of the most exciting fields. While it involves item detection in surveillance structures, speed plays a critical role in locating an item fast and alerting the expert. This effort tried to reap the similar and its capable of producing a quicker result compared to the previously current structures. The longer-time period effort of the future device is to extend a more quantity of types of weapons and classify them. The correctness of the weapon noticed can be advanced by the use of extraordinary forms of algorithms. A likely way to enhance this painting is to discover a hidden weapon that cannot be detected by the usage of the regular digicam. Also, analyzing the behavior of the humans to locate any suspicious sports like hiding the weapon can be completed to improve this surveillance machine. The alert machine also can be stepped forward to inform a couple of users if the weapon is detected. A surveillance gadget with these features may be useful to prevent violent crimes and provide protection to the public.

References

1. Raturi, G.; Rani, P.; Madan, S.; Dosanjh, S. ADoCW: An Automated Method for Detection of Concealed Weapon. In Proceedings of the Fifth International Conference on Image Information Processing, Shimla, India, 15–17 November 2019; pp. 181–186.
2. Bhagyalakshmi, P.; Indhumathi, P.; Lakshmi, R.; Bhavadharini, D. Real-Time Video Surveillance for Automated Weapon Detection. *Int. J. Trend Sci. Res. Dev.* 2019, 3, 465–470. [CrossRef]
3. Lim, J.; Jobayer, M.I.A.; Baskaran, V.M.; Lim, J.M.; Wong, K.; See, J. Gun Detection in Surveillance Videos using Deep Neural Networks. In Proceedings of the Asia-Pacific Signal and Information Processing Association Annual Summit and Conference, Lanzhou, China, 18–21 November 2019; pp. 1998–2002. [CrossRef]
4. Yuan, J.; Guo, C. A deep learning method for detection of dangerous equipment. In Proceedings of the Eighth International Conference on Information Science and Technology, Granada, Cordoba, and Seville, Spain, 30 June–6 July 2018; pp. 159–164. [CrossRef]
5. Romero, D.; Salamea, C. Convolutional models for the detection of firearms in surveillance videos. *Appl. Sci.* 2019, 9, 2965. [CrossRef]
6. Ilgin, F.Y. Energy-based spectrum sensing with copulas for cognitive radios. *Bull. Polish Acad. Sci. Tech. Sci.* 2020, 68, 829–834. [CrossRef]
7. Navalgund, U.V.; Priyadharshini, K. Crime Intention Detection System Using Deep Learning. In Proceedings of the International Conference on Circuits and Systems in Digital Enterprise Technology, Kottayam, India, 21–22 December 2018; pp. 1–6. [CrossRef]
8. Chandan, G.; Jain, A.; Jain, H. Real-time object detection and tracking using Deep Learning and OpenCV. In Proceedings of the International Conference on Inventive Research in Computing Applications, Coimbatore, India, 11–12 July 2018; pp. 1305–1308. [CrossRef]
9. Deng, L.; Yu, D. Deep learning: Methods and applications. *Found. Trends Signal Process.* 2014, 7, 197–387. [CrossRef]
10. Bengio, Y. Learning deep architectures for AI. *Found. Trends Mach. Learn.* 2009, 2, 1–27. [CrossRef]
11. Lecun, Y.; Bengio, Y.; Hinton, G. Deep learning. *Nature* 2015, 521, 436–444. [CrossRef] [PubMed]
12. Song, H.A.; Lee, S.Y. Hierarchical representation using NMF. *Lect. Notes Comput. Sci.* 2013, 8226, 466–473. [CrossRef]
11. “Detection and Classification of Different Weapon Types Using Deep Learning” Volkan Kaya 1,*, Servet Tuncer 2 and Ahmet Baran 1 Department of Computer Engineering, Erzincan Binali Yildirim University, Erzincan 24100, Turkey; baran@erzincan.edu.tr 2 Department of Electrical and Electronic Engineering, Firat University, Elazığ 23119, Turkey; stuncer@firat.edu.tr