

Weapon Detection Using Deep Learning: A Systematic Review

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ABSTRACT

Weapon detection is a time-critical vision task spanning public-space CCTV, body-worn cameras, and UAVs. This work presents a systematic review of deep learning approaches across visible-spectrum modalities, covering model families, datasets, evaluation protocols, and deployment constraints. We find that YOLO-family one-stage detectors dominate visible-spectrum applications by offering the best accuracy–latency trade-off for real-time inference, while X-ray research leverages large public benchmarks (e.g., SIXRay, OPIXray, PIDray) and increasingly adopts attention and anchor-free designs to handle clutter, overlap, and deformation. Performance is highly sensitive to scale-aware preprocessing, dataset realism/diversity, and, for video, temporal smoothing/aggregation. Persistent gaps include cross-site generalization, small/occluded weapon detection, rigorous calibration/uncertainty, and responsible deployment (privacy, bias, adversarial robustness). We distill best practices—compact YOLO variants with multi-scale heads and optional tiling; brightness/contrast normalization; synthetic+real data mixing; pose/orientation cues when architecture-compatible—and outline opportunities in foundation-model features, multimodal fusion (RGB–X-ray/IR), self-/active learning for long-tail classes, and standardized benchmarking with consistent mAP/F1/FPS reporting and small-object stratification..

Keywords: Weapon detection, object detection, CCTV, Deep Learning, YOLO, Real-time vision.

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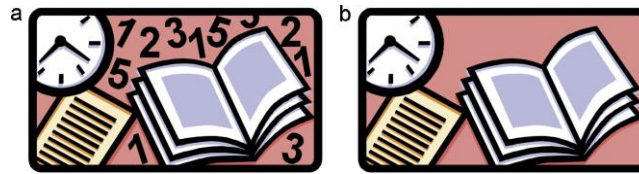


Fig. 1 - (a) first picture; (b) second picture.

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References

1. A. Warsi, M. Abdullah, M. N. Husen, and M. Yahya, "Automatic handgun and knife detection algorithms: A review," in Proc. 14th Int. Conf. Ubiquitous Information Management and Communication (IMCOM), IEEE, 2020, pp. 1–9, doi: 10.1109/IMCOM48794.2020.9001725.
2. R. Debnath and M. K. Bhowmik, "A comprehensive survey on computer vision based concepts, methodologies, analysis and applications for automatic gun/knife detection," J. Vis. Commun. Image Represent., vol. 78, p. 103165, 2021, doi: 10.1016/j.jvcir.2021.103165.
3. A. AlZaabi, M. A. Talib, A. B. Nassif, A. Sajwani, and O. Einea, "A systematic literature review on machine learning in object detection security," in Proc. 2020 IEEE 5th Int. Conf. Computing Communication and Automation (ICCCA), 2020, pp. 136–139, doi: 10.1109/ICCCA49541.2020.9250836.

4. Z.-Q. Zhao, P. Zheng, S.-T. Xu, and X. Wu, "Object detection with deep learning: A review," *IEEE Trans. Neural Netw. Learn. Syst.*, vol. 30, no. 11, pp. 3212–3232, 2019, doi: 10.1109/TNNLS.2018.2876865.
5. Y. Xiao, Z. Tian, J. Yu, Y. Zhang, S. Liu, S. Du, and X. Lan, "A review of object detection based on deep learning," *Multimedia Tools Appl.*, vol. 79, nos. 33–34, pp. 23729–23791, 2020, doi: 10.1007/s11042-020-08976-6.
6. K. Tong, Y. Wu, and F. Zhou, "Recent advances in small object detection based on deep learning: A review," *Image Vis. Comput.*, vol. 97, p. 103910, 2020, doi: 10.1016/j.imavis.2020.103910.
7. V. Sharma and R. N. Mir, "A comprehensive and systematic look up into deep learning based object detection techniques: A review," *Comput. Sci. Rev.*, vol. 38, p. 100301, 2020, doi: 10.1016/j.cosrev.2020.100301.
8. L. Aziz, M. S. B. H. Salam, U. U. Sheikh, and S. Ayub, "Exploring deep learning-based architecture, strategies, applications and current trends in generic object detection: A comprehensive review," *IEEE Access*, vol. 8, pp. 170461–170495, 2020, doi: 10.1109/ACCESS.2020.3021508.
9. S. Bouraya and A. Belangour, "Deep learning based neck models for object detection: A review and a benchmarking study," *Int. J. Adv. Comput. Sci. Appl.*, vol. 12, no. 11, 2021, doi: 10.14569/IJACSA.2021.0121119.
10. M. Ahmed, K. A. Hashmi, A. Pagani, M. Liwicki, D. Stricker, and M. Z. Afzal, "Survey and performance analysis of deep learning based object detection in challenging environments," *Sensors*, vol. 21, no. 15, p. 5116, 2021, doi: 10.3390/s21155116.
11. M. M. Fernandez-Carrobles, O. Deniz, and F. Maroto, "Gun and knife detection based on faster R-CNN for video surveillance," in *Proc.*, 2019, pp. 441–452. doi: 10.1007/978-3-030-31321-0_38. Available: http://link.springer.com/10.1007/978-3-030-31321-0_38
12. N. Vallez, A. Velasco-Mata, J. J. Corroto, and O. Deniz, "Weapon detection for particular scenarios using deep learning," in *Proc.*, 2019, pp. 371–382. doi: 10.1007/978-3-030-31321-0_32. Available: http://link.springer.com/10.1007/978-3-030-31321-0_32
13. A. Castillo, S. Tabik, F. Pérez, R. Olmos, and F. Herrera, "Brightness guided preprocessing for automatic cold steel weapon detection in surveillance videos with deep learning," *Neurocomputing*, vol. 330, pp. 151–161, 2019. doi: 10.1016/j.neucom.2018.10.076. Available: <https://linkinghub.elsevier.com/retrieve/pii/S0925231218313365>
14. S. Loganathan, G. Kariyawasam, and P. Sumathipala, "Suspicious activity detection in surveillance footage," in *Proc. 2019 Int. Conf. Electrical and Computing Technologies and Applications (ICECTA)*, 2019, pp. 1–4. doi: 10.1109/ICECTA48151.2019.8959600. Available: <https://ieeexplore.ieee.org/document/8959600/>
15. J. Suarez-Paez, M. Salcedo-Gonzalez, A. Climente, M. Esteve, J. A. Gómez, C. E. Palau, and I. Pérez-Llopis, "A novel low processing time system for criminal activities detection applied to command and control citizen security centers," *Information*, vol. 10, no. 12, p. 365, 2019.
16. F. Pérez-Hernández, S. Tabik, A. Lamas, R. Olmos, H. Fujita, and F. Herrera, "Object detection binary classifiers methodology based on deep learning to identify small objects handled similarly: Application in video surveillance," *Knowledge-Based Systems*, vol. 194, p. 105590, 2020. doi: 10.1016/j.knsys.2020.105590. Available: <https://linkinghub.elsevier.com/retrieve/pii/S0950705120300678>

17. A. Egiazarov, F. M. Zennaro, and V. Mavroeidis, "Firearm detection via convolutional neural networks: Comparing a semantic segmentation model against end-to-end solutions," in Proc. 2020 IEEE Int. Conf. Big Data (Big Data), 2020, pp. 1796–1804. doi: 10.1109/BigData50022.2020.9377745. Available: <https://ieeexplore.ieee.org/document/9377745/>
18. H. Peng, Y. Zhang, S. Yang, and B. Song, "Battlefield image situational awareness application based on deep learning," IEEE Intelligent Systems, vol. 35, no. 1, pp. 36–43, 2020. doi: 10.1109/MIS.2019.2953685. Available: <https://ieeexplore.ieee.org/document/8903625/>
19. J. Rose, T. Bourlai, and J. A. Loudermilk, "Assessment of data augmentation techniques for firearm detection in surveillance videos," in Proc. 2020 IEEE Int. Conf. Big Data (Big Data), 2020, pp. 1838–1846. doi: 10.1109/BigData50022.2020.9378267. Available: <https://ieeexplore.ieee.org/document/9378267/>
20. M. K. Galab, A. Taha, and H. H. Zayed, "Adaptive technique for brightness enhancement of automated knife detection in surveillance video with deep learning," Arabian Journal for Science and Engineering, vol. 46, no. 4, pp. 4049–4058, 2021. doi: 10.1007/s13369-021-05401-4. Available: <http://link.springer.com/10.1007/s13369-021-05401-4>
21. V. Kaya, S. Tuncer, and A. Baran, "Detection and classification of different weapon types using deep learning," Applied Sciences, vol. 11, no. 16, p. 7535, 2021. doi: 10.3390/app11167535. Available: <https://www.mdpi.com/2076-3417/11/16/7535>
22. M. T. Bhatti, M. G. Khan, M. Aslam, and M. J. Fiaz, "Weapon detection in real-time CCTV videos using deep learning," IEEE Access, vol. 9, pp. 34366–34382, 2021. doi: 10.1109/ACCESS.2021.3059170. Available: <https://ieeexplore.ieee.org/document/9353483/>
23. T. S. S. Hashmi, N. U. Haq, M. M. Fraz, and M. Shahzad, "Application of deep learning for weapons detection in surveillance videos," in Proc. 2021 Int. Conf. Digital Futures and Transformative Technologies (ICoDT2), 2021, pp. 1–6. doi: 10.1109/ICoDT252288.2021.9441523. Available: <https://ieeexplore.ieee.org/document/9441523/>
24. N. U. L. Haq, T. S. S. Hashmi, M. M. Fraz, and M. Shahzad, "Rotation aware object detection model with applications to weapons spotting in surveillance videos," in Proc. 2021 Int. Conf. Digital Futures and Transformative Technologies (ICoDT2), 2021, pp. 1–6. doi: 10.1109/ICoDT252288.2021.9441538. Available: <https://ieeexplore.ieee.org/document/9441538/>
25. A. Velasco-Mata, J. Ruiz-Santaquiteria, N. Vallez, and O. Deniz, "Using human pose information for handgun detection," Neural Computing and Applications, vol. 33, no. 24, pp. 17273–17286, 2021. doi: 10.1007/s00521-021-06317-8. Available: <https://link.springer.com/10.1007/s00521-021-06317-8>
26. J. Salido, V. Lomas, J. Ruiz-Santaquiteria, and O. Deniz, "Automatic handgun detection with deep learning in video surveillance images," Applied Sciences, vol. 11, no. 13, p. 6085, 2021. doi: 10.3390/app11136085. Available: <https://www.mdpi.com/2076-3417/11/13/6085>
27. S. Narejo, B. Pandey, D. Esenarro Vargas, C. Rodriguez, and M. R. Anjum, "Weapon detection using YOLO V3 for smart surveillance system," Mathematical Problems in Engineering, vol. 2021, pp. 1–9, 2021. doi: 10.1155/2021/9975700. Available: <https://www.hindawi.com/journals/mpe/2021/9975700/>
28. J. Y. Lim, M. I. A. Jobayer, V. M. Baskaran, J. M. Y. Lim, J. See, and K.-S. Wong, "Deep multi-level feature pyramids: Application for non-canonical firearm detection in video surveillance," Engineering Applications of Artificial Intelligence, vol. 97, p. 104094, 2021. doi:

- 10.1016/j.engappai.2020.104094. Available:
<https://linkinghub.elsevier.com/retrieve/pii/S0952197620303456>
29. A. A. Ahmed and M. Echi, "Hawk-eye: An AI-powered threat detector for intelligent surveillance cameras," *IEEE Access*, vol. 9, pp. 63283–63293, 2021. doi: 10.1109/ACCESS.2021.3074319. Available: <https://ieeexplore.ieee.org/document/9408578/>
30. L. Kong, J. Wang, and P. Zhao, "YOLO-g: A lightweight network model for improving the performance of military targets detection," *IEEE Access*, vol. 10, pp. 55546–55564, 2022.
31. J. A. Kumar et al., "Weapon Detection Identification and Classification for DCNN and YOLO-V5 Techniques," 2024 International Conference on IoT Based Control Networks and Intelligent Systems (ICICNIS), Bengaluru, India, 2024, pp. 1317-1323, doi: 10.1109/ICICNIS64247.2024.10823365.
32. Shanthi, P., and V. Manjula. "A systematic review on CNN-YOLO techniques for face and weapon detection in crime prevention." *Discover Computing* 28.1 (2025): 204.
33. D. Chitravanshi, A. Malik, H. Saini, S. Avasthi, K. Agarwal and Y. Grover, "Weapon Detection from Images using YOLO and OpenCV," 2024 First International Conference on Technological Innovations and Advance Computing (TIACOMP), Bali, Indonesia, 2024, pp. 560-565, doi: 10.1109/TIACOMP64125.2024.00098.
34. Yadav, Rajeshwar, et al. "SIRD-YOLO: an enhanced deep learning model for weapon detection using spatial interactions and diverse receptive fields." *Innovations in Systems and Software Engineering* (2024): 1-17.