

## 7. References

- Al-Sa'd, M., Kiranyaz, S., Ahmad, I., Sundell, C., Vakkuri, M., & Gabbouj, M. (2022). A Social Distance Estimation and Crowd Monitoring System for Surveillance Cameras. *Sensors*, 22(2). <https://doi.org/10.3390/s22020418>
- Alzahrani, B., Barnawi, A., Irshad, A., Alhothali, A., Alotaibi, R., & Shafiq, M. (2022). A secure key agreement scheme for unmanned aerial vehicles-based crowd monitoring system. *Computers, Materials and Continua*, 70(3). <https://doi.org/10.32604/cmc.2022.020774>
- Castellano, G., Castiello, C., Mencar, C., & Vessio, G. (2020). Crowd Detection in Aerial Images Using Spatial Graphs and Fully-Convolutional Neural Networks. *IEEE Access*, 8. <https://doi.org/10.1109/ACCESS.2020.2984768>
- Castellano, G., Cotardo, E., Mencar, C., & Vessio, G. (2023). Density-based clustering with fully-convolutional networks for crowd flow detection from drones. *Neurocomputing*, 526. <https://doi.org/10.1016/j.neucom.2023.01.059>
- Chebil, K., Htiouech, S., & Khemakhem, M. (2022). Toward Optimal Periodic Crowd Tracking via Unmanned Aerial Vehicles. *SSRN Electronic Journal*. <https://doi.org/10.2139/ssrn.4100367>
- Chen, G., Wang, W., He, Z., Wang, L., Yuan, Y., Zhang, D., Zhang, J., Zhu, P., Van Gool, L., Han, J., Hoi, S., Hu, Q., Liu, M., Sciarrone, A., Sun, C., Garibotto, C., Tran, D. N. N., Lavagetto, F., Haleem, H., ... Luo, Z. (2021). VisDrone-MOT2021: The Vision Meets Drone Multiple Object Tracking Challenge Results. *Proceedings of the IEEE International Conference on Computer Vision, 2021-October*. <https://doi.org/10.1109/ICCVW54120.2021.00318>
- Cocca, U., Giusti, A., & Roveda, L. (2022). Hand Gestures-Based Smooth 3D Trajectories Computation Applied to Real-Time Drone Control by Tracking 2D Hand Landmarks. *International Conference on Electrical, Computer, Communications and Mechatronics Engineering, ICECCME 2022*. <https://doi.org/10.1109/ICECCME55909.2022.9987798>
- Fadzil, N., Abu Bakar, N. H., Idrus, S. M., Azmi, A. I., Mohammad, S. H., & Ali, N. (2021). Recent Development of Crowd Monitoring Technology Solution for Covid-19 Prevention at Airport Terminal. In *International Journal of Nanoelectronics and Materials* (Vol. 14, Issue Special Issue InCAPE).
- Fan, Z., Zhang, H., Zhang, Z., Lu, G., Zhang, Y., & Wang, Y. (2022). A survey of crowd counting and density estimation based on convolutional neural network. *Neurocomputing*, 472. <https://doi.org/10.1016/j.neucom.2021.02.103>
- Gajjar, V., Khandhediya, Y., & Gurnani, A. (2017). Human detection and tracking for video surveillance: A cognitive science approach. *Proceedings - 2017 IEEE International Conference on Computer Vision Workshops, ICCVW 2017, 2018-January*. <https://doi.org/10.1109/ICCVW.2017.330>
- Jiang, M., Lin, J., & Jane Wang, Z. (2021). SHUFFLECOUNT: TASK-SPECIFIC KNOWLEDGE DISTILLATION FOR CROWD COUNTING. *Proceedings - International Conference on Image Processing, ICIP, 2021-September*. <https://doi.org/10.1109/ICIP42928.2021.9506698>
- Jiang, M., Lin, J., & Wang, Z. J. (2021). A smartly simple way for joint crowd counting and localization. *Neurocomputing*, 459. <https://doi.org/10.1016/j.neucom.2021.06.055>
- Lan, W., Dang, J., Wang, Y., & Wang, S. (2018). Pedestrian detection based on yolo network model. *Proceedings of 2018 IEEE International Conference on Mechatronics and Automation, ICMA 2018*. <https://doi.org/10.1109/ICMA.2018.8484698>
- Liang, D., Chen, X., Xu, W., Zhou, Y., & Bai, X. (2022). TransCrowd: weakly-supervised crowd counting with transformers. *Science China Information Sciences*, 65(6). <https://doi.org/10.1007/s11432-021-3445-y>
- Molchanov, V. V., Vishnyakov, B. V., Vizilter, Y. V., Vishnyakova, O. V., & Knyaz, V. A. (2017). Pedestrian detection in video surveillance using fully convolutional YOLO neural network. *Automated Visual Inspection and Machine Vision II*, 10334. <https://doi.org/10.1117/12.2270326>
- Nadeem, A., Ashraf, M., Qadeer, N., Rizwan, K., Mehmood, A., AlZahrani, A., Noor, F., & Abbasi, Q. H. (2022). Tracking Missing Person in Large Crowd Gathering Using Intelligent Video Surveillance. *Sensors*, 22(14). <https://doi.org/10.3390/s22145270>
- Shadakshri, V. H. C., Veena, M. B., & Dev, V. K. R. G. (2022). OpenCV Implementation of Grid-based Vertical Safe Landing for UAV using YOLOv5. *International Journal of Advanced Computer Science and Applications*, 13(9). <https://doi.org/10.14569/IJACSA.2022.0130957>
- Sindagi, V. A., & Patel, V. M. (2017). CNN-Based cascaded multi-task learning of high-level prior and density estimation for crowd counting. *2017 14th IEEE International Conference on Advanced Video and Signal Based Surveillance, AVSS 2017*. <https://doi.org/10.1109/AVSS.2017.8078491>

- Syed Ameer Abbas, S., Oliver Jayaprakash, P., Anitha, M., & Vinitha Jaini, X. (2018). Crowd detection and management using cascade classifier on ARMv8 and OpenCV-Python. *Proceedings of 2017 International Conference on Innovations in Information, Embedded and Communication Systems, ICIIECS 2017, 2018-January*. <https://doi.org/10.1109/ICIIECS.2017.8275988>
- Vaswani, A., Shazeer, N., Parmar, N., Uszkoreit, J., Jones, L., Gomez, A. N., Kaiser, Ł., & Polosukhin, I. (2017). An image is worth 16\*16 words: transformers for image recognition at scale. *Advances in Neural Information Processing Systems, 2017-Decem*.
- Wang, X., Lv, J., & Yun, Z. (2022). A Real-time improved pedestrian dead reckoning trajectory tracking algorithm. *International Archives of the Photogrammetry, Remote Sensing and Spatial Information Sciences - ISPRS Archives, 46(3/W1-2022)*. <https://doi.org/10.5194/isprs-archives-XLVI-3-W1-2022-197-2022>
- Wen, L., Du, D., Zhu, P., Hu, Q., Wang, Q., Bo, L., & Lyu, S. (2021). Detection, Tracking, and Counting Meets Drones in Crowds: A Benchmark. *Proceedings of the IEEE Computer Society Conference on Computer Vision and Pattern Recognition*. <https://doi.org/10.1109/CVPR46437.2021.00772>
- Wu, Y., Sui, Y., & Wang, G. (2017). Vision-Based Real-Time Aerial Object Localization and Tracking for UAV Sensing System. *IEEE Access, 5*. <https://doi.org/10.1109/ACCESS.2017.2764419>
- Xiao, Y., Kamat, V. R., & Menassa, C. C. (2019). Human tracking from single RGB-D camera using online learning. *Image and Vision Computing, 88*. <https://doi.org/10.1016/j.imavis.2019.05.003>
- Yadav, S., Gulia, P., Gill, N. S., & Chatterjee, J. M. (2022). A Real-Time Crowd Monitoring and Management System for Social Distance Classification and Healthcare Using Deep Learning. *Journal of Healthcare Engineering, 2022*. <https://doi.org/10.1155/2022/2130172>
- Yan, Z., Duckett, T., & Bellotto, N. (2020). Online learning for 3D LiDAR-based human detection: experimental analysis of point cloud clustering and classification methods. *Autonomous Robots, 44(2)*. <https://doi.org/10.1007/s10514-019-09883-y>
- Zhang, C., Yang, Z., Liao, L., You, Y., Sui, Y., & Zhu, T. (2022). RPEOD: A Real-Time Pose Estimation and Object Detection System for Aerial Robot Target Tracking. *Machines, 10(3)*. <https://doi.org/10.3390/machines10030181>
- Zhang, G., Pan, Y., Zhang, L., & Tiong, R. L. K. (2020). Cross-scale generative adversarial network for crowd density estimation from images. *Engineering Applications of Artificial Intelligence, 94*. <https://doi.org/10.1016/j.engappai.2020.103777>
- Zhu, L., Li, C., Yang, Z., Yuan, K., & Wang, S. (2020). Crowd density estimation based on classification activation map and patch density level. *Neural Computing and Applications, 32(9)*. <https://doi.org/10.1007/s00521-018-3954-7>