

Tracking and Computation of Characteristics of the Movement of People in Groups on Video Using Convolutional Neural Networks

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Abstract

This paper proposes an approach for tracking the behavior of people in a group on video by using convolutional neural networks. At the beginning, definitions of group movement of people are given, and features for accompaniment are defined that can be used to analyze people's behavior. Next, an algorithm is proposed for calculating the distance between people in video, which includes three stages: detection and tracking of objects, coordinate transformation, calculation of the distance between people and detection of distance violations. The results of experimental studies and comparison with known algorithms are presented, which confirms the effectiveness of the algorithm.

REFERENCES

1. Multiple Object Tracking Benchmark. <https://motchallenge.net/>. Accessed June 24, 2024.
2. Ciaparrone, G., Sánchez, F.L., Tabik, S., Troiano, L., Tagliaferri, R., and Herrera, F., Deep learning in video multi-object tracking:

A survey, *Neurocomputing*, 2020, vol. 381, pp. 61–88. <https://doi.org/10.1016/j.neucom.2019.11.023>

Article MATH Google Scholar

3. Charles, P., Jasmine Sultana, S., Hemalatha, P., and Keerti, E., Multiple person detection and tracking using Convolutional Neural Network, *Int. J. Adv. Res. Innovation*, 2020, vol. 8, no. 2, pp. 156–159.

Google Scholar

4. Iqbal, U., Milan, A., and Gall, J., Pose Track: Joint Multi-person Pose Estimation and Tracking, *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR)*, 2017, pp. 4654–4663.
5. Singh, K., Rajora, S., Vishwakarma, D.K., Tripathi G., Kumar, S., and Walia, G.S., Crowd anomaly detection using aggregation of ensembles of fine-tuned ConvNets, *Neurocomputing*, 2020, vol. 371, pp. 188–198. <https://doi.org/10.1016/j.neucom.2019.08.059>

Article MATH Google Scholar

6. Zhang, L. and Han, J., Recognition of abnormal behavior of crowd based on spatial location feature, *2020 IEEE 9th Joint International Information Technology and Artificial Intelligence Conference (ITAIC)*, 2020, pp 736–741. <https://doi.org/10.1109/ITAIC49862.2020.9338944>
7. Amnah Aldayri and Waleed Albattah, Taxonomy of anomaly detection techniques in crowd scenes, *Sensors*, 2022, vol. 22, no. 16, p. 6080. <https://doi.org/10.3390/s22166080>

Article MATH Google Scholar

8. Abdullah, F. and Jalal, A., Multi-pedestrians anomaly detection via conditional random field and deep learning, *4th International Conference on Advancements in Computational Sciences (ICACS)*, 2023, pp. 1–6. <https://doi.org/10.1109/ICACS55311.2023.10089730>
9. Li, Y.C., Jia, R.S., Hu, Y.X., Han, D.N., and Sun, H.M., Crowd density estimation based on multi scale features fusion network with reverse attention mechanism, *Appl. Intell.*, 2022, vol. 52, pp. 13 097–13 113. <https://doi.org/10.1007/s10489-022-03187-y>

Article Google Scholar

10. Fan, Z., Zhang, H., Zhang, Z., Lu, G., Zhang, Y., and Wang, Y., A survey of crowd counting and density estimation based on convolutional neural network, *Neurocomputing*, 2022, vol. 472, pp. 224–251. <https://doi.org/10.1016/j.neucom.2021.02.103>

Article MATH Google Scholar

11. Alrowais, F., Alotaibi, S.S., Al-Wesabi, F.N., Negm, N., Alabdan, R., Marzouk, R., Mehanna, A.S., and Al Duhayyim, M., Deep transfer learning enabled intelligent object detection for crowd density analysis on video surveillance systems, *Appl. Sci.*, 2022, vol. 12, p. 6665. <https://doi.org/10.3390/app12136665>

Article Google Scholar

12. Elbishlawi, S., Abdelpakey, M.H., Eltantawy, A., Shehata, M.S., and Mohamed, M.M., Deep learning-based crowd scene analysis survey, *J. Imaging*, 2020, vol. 6, p. 95. <https://doi.org/10.3390/jimaging6090095>

Article Google Scholar

13. Bendali-Braham M. et al., Recent trends in crowd analysis: a review, *Mach. Learn. Appl.*, 2021, vol. 4, 100023. <https://doi.org/10.1016/j.mlwa.2021.100023>

Article MATH Google Scholar

14. Chaudhary, D., Kumar, S., and Dhaka, V.S., Video based human crowd analysis using machine learning: A survey, *Comput. Methods Biomechan. Biomed. Eng.: Imaging Visualization*, 2021, vol. 10, no. 2, pp. 113–131. <https://doi.org/10.1080/21681163.2021.1986859>

Article MATH Google Scholar

15. Kumar, A. and Arunnehru, J., Crowd behavior monitoring and analysis in surveillance applications: A survey, *Turk. J. Comput. Math. Educ.*, 2021, vol. 12, no. 7, pp. 2322–2336.

MATH Google Scholar

16. Ye, S., Bohush, R., Chen, C., Zakharova, I., and Ablameyko, S., Person tracking and re-identification in video for indoor

multi-camera surveillance systems, *Pattern Recognit. Image Anal.*, 2020, vol. 30, no. 4, pp. 827–837. <https://doi.org/10.1134/S1054661820040136>

Article Google Scholar

17. Shah, J., Chandaliya, M., Bhuta, H., and Kanani, P., Social distancing detection using computer vision, *5th International Conference on Computing Methodologies and Communication (ICCMC)*, 2021, pp. 1359–1365. <https://doi.org/10.1109/ICCMC51019.2021.9418312>

18. Gündüz, M. and Işık, G., A new YOLO-based method for social distancing from real-time videos, *Neural Comput. Appl.*, 2023, vol. 35 pp. 15 261–15 271. <https://doi.org/10.1007/s00521-023-08556-3>

Article MATH Google Scholar

19. Ahmed, I., Ahmad, M., Rodrigues, J., Jeon, G., and Din, S., A deep learning-based social distance monitoring framework for COVID-19, *Sustainable Cities Soc.*, 2021, vol. 65. 102571. <https://doi.org/10.1016/j.scs.2020.102571>

Article Google Scholar

20. Bernasco, W., Hoeben, E., Koelma, D., Liebst, L., Thomas, J., Appelman, J., Snoek, C., and Lindegaard, M., Promise into practice: Application of computer vision in empirical research on social distancing, *Sociol. Methods Res.*, 2023, vol. 52, no. 3, pp.1239–1287. <https://doi.org/10.1177/00491241221099>

Article MathSciNet Google Scholar

21. Saponara, S., Elhanashi, A., and Qinghe, Z., Developing a real-time social distancing detection system based on YOLOv4-tiny and bird-eye view for COVID-19, *J. Real-Time Image Process.*, 2022, vol.19. pp. 551–563. <https://doi.org/10.1007/s11554-022-01203-5>

Article Google Scholar

22. Rahim, A., Maqbool, A., and Rana, T., Monitoring social distancing under various low light conditions with deep learning and a single motionless time of flight camera, *PLOS One*, 2021, vol. 16. e0247440. <https://doi.org/10.1371/journal.pone.0247440>

Article Google Scholar

23. Oxford Town Centre Dataset.
<https://www.kaggle.com/datasets/almightyj/oxford-town-centre>. Accessed May 01, 2024.
24. Krytsky, A., Detection and tracking of objects in the problem of monitoring social distance, *Master's Diss.*, BSU, Minsk, 2024.
25. Das, S. et al., Computer vision-based social distancing surveillance solution with optional automated camera calibration for large scale deployment. arXiv preprint arXiv:2104.10891 (2021). <https://doi.org/10.48550/arXiv.2104.10891>
26. Sangeetha, R.G. and Jaya Aravindh, V.V., Social distance detection using deep learning and risk management system, 2023.
arXiv:2304.10259. <https://doi.org/10.48550/arXiv.2304.10259>

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Ethics declarations

The authors of this work declare that they have no conflicts of interest.

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- **calculation of the distance between people**
- **video surveillance**