REFERENCES

- C. Wang, Y. Wang, M. Xu, and D. J. Crandall, "Stepwise goal-driven networks for trajectory prediction," *IEEE Robotics and Automation Letters*, vol. 7, no. 2, pp. 2716–2723, 2022.
- [2] K. Mangalam, Y. An, H. Girase, and J. Malik, "From goals, waypoints & paths to long term human trajectory forecasting," in 2021 IEEE/CVF International Conference on Computer Vision (ICCV), 2021, pp. 15213– 15222.
- [3] Y. Zhang, C. Wang, X. Wang, W. Zeng, and W. Liu, "FairMOT: On the fairness of detection and re-identification in multiple object tracking," *International Journal of Computer Vision*, vol. 129, no. 11, pp. 3069–3087, sep 2021. [Online]. Available: https://doi.org/10.1007%2Fs11263-021-01513-4
- [4] K. Bimbraw, "Autonomous cars: Past, present and future a review of the developments in the last century, the present scenario and the expected future of autonomous vehicle technology," in 2015 12th International Conference on Informatics in Control, Automation and Robotics (ICINCO), vol. 01, 2015, pp. 191–198.
- [5] F. Schneemann and P. Heinemann, "Context-based detection of pedestrian crossing intention for autonomous driving in urban environments," in 2016 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS), 2016, pp. 2243–2248.
- [6] A. Kendall, Y. Gal, and R. Cipolla, "Multi-task learning using uncertainty to weigh losses for scene geometry and semantics," 2017. [Online]. Available: https://arxiv.org/abs/1705.07115
- [7] C.-Y. Wang, A. Bochkovskiy, and H.-Y. M. Liao, "Yolov7: Trainable bag-of-freebies sets new state-of-the-art for real-time object detectors," 2022. [Online]. Available: https://arxiv.org/abs/2207.02696
- [8] X. Shao, J. Wei, D. Guo, R. Zheng, X. Nie, G. Wang, and Y. Zhao, "Pedestrian detection algorithm based on improved faster rcnn," in 2021 IEEE 5th Advanced Information Technology, Electronic and Automation Control Conference (IAEAC), vol. 5, 2021, pp. 1368–1372.
- [9] T.-Y. Lin, P. Dollár, R. Girshick, K. He, B. Hariharan, and S. Belongie, "Feature pyramid networks for object detection," 2016. [Online]. Available: https://arxiv.org/abs/1612.03144
- [10] O. Kesa, O. Styles, and V. Sanchez, "Multiple object tracking and forecasting: Jointly predicting current and future object locations," in 2022 IEEE/CVF Winter Conference on Applications of Computer Vision Workshops (WACVW), 2022, pp. 560–569.
- [11] Z. Wang, L. Zheng, Y. Liu, Y. Li, and S. Wang, "Towards real-time multi-object tracking," 2019. [Online]. Available: https://arxiv.org/abs/1909.12605
- [12] N. Wojke, A. Bewley, and D. Paulus, "Simple online and realtime tracking with a deep association metric," 2017. [Online]. Available: https://arxiv.org/abs/1703.07402
- [13] Y. Du, Y. Song, B. Yang, and Y. Zhao, "Strongsort: Make deepsort great again," 2022. [Online]. Available: https://arxiv.org/abs/2202.13514
- [14] K. Mangalam, H. Girase, S. Agarwal, K.-H. Lee, E. Adeli, J. Malik, and A. Gaidon, "It is not the journey but the destination: Endpoint conditioned trajectory prediction," 2020. [Online]. Available: https://arxiv.org/abs/2004.02025
- [15] T. Salzmann, B. Ivanovic, P. Chakravarty, and M. Pavone, "Trajectron++: Dynamically-feasible trajectory forecasting with heterogeneous data," 2020. [Online]. Available: https://arxiv.org/abs/2001.03093
- [16] A. Rasouli, I. Kotseruba, T. Kunic, and J. Tsotsos, "Pie: A large-scale dataset and models for pedestrian intention estimation and trajectory prediction," in 2019 IEEE/CVF International Conference on Computer Vision (ICCV), 2019, pp. 6261–6270.