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Cloud and Edge Computing for Connected and Automated Vehicles

Qi Zhu

Northwestern University
qzhu@northwestern.edu

Bo Yu

PerceptIn
bo.yu@perceptin.io

Ziran Wang

Purdue University
ryanwang11@hotmail.com

Jie Tang

South China University of Technology
cstangjie@scut.edu.cn

Qi Alfred Chen

University of California, Irvine
alfchen@uci.edu

Zihao Li

Purdue University
zihao@purdue.edu

Xiangguo Liu

Northwestern University
XiangguoLiu2023@u.northwestern.edu

Yunpeng Luo

University of California, Irvine
yunpel3@uci.edu

Lingzi Tu

Southwest Jiaotong University
linzitu@outlook.com



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now Publishers Inc.
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Tel. +31-6-51115274

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Cloud and Edge Computing for Connected and Automated Vehicles

Qi Zhu¹, Bo Yu², Ziran Wang³, Jie Tang⁴, Qi Alfred Chen⁵, Zihao Li³, Xiangguo Liu¹, Yunpeng Luo⁵ and Lingzi Tu⁶

¹Northwestern University, USA; qzhu@northwestern.edu, XiangguoLiu2023@u.northwestern.edu

²PerceptIn, USA; bo.yu@perception.io

³Purdue University, USA; ryanwang11@hotmail.com, zihao@purdue.edu

⁴South China University of Technology, China; cstangjie@scut.edu.cn

⁵University of California, Irvine, USA; alfchen@uci.edu, yunpel3@uci.edu

⁶Southwest Jiaotong University, China; linzitu@outlook.com

ABSTRACT

The recent development of cloud computing and edge computing shows great promise for the Connected and Automated Vehicle (CAV), by enabling CAVs to offload their massive on-board data and heavy computing tasks. Leveraging the Internet of Things (IoT) technology, different entities in the intelligent transportation system (e.g., vehicles, infrastructure, traffic management centers, etc.) get connected with each other, thus making the entire system smarter, faster, and more efficient. However, these advances also bring significant challenges to public authorities, industry, as well as scientific communities. In terms of system design and control, current cloud and edge architecture of CAVs need to be refined or even redesigned to better function under uncertainties in demand, and to better cooperate with

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existing conventional vehicles and infrastructure. From the performance assessment perspective, models and simulation tools based on artificial intelligence and big data have been widely developed for validation and evaluation of cloud computing and edge computing, but the validity of these models needs to be re-examined with field implementations.

Finally, while the increasing connectivity among vehicles and infrastructures may help improve their perception of the environment and enable coordinated decision making, it also presents new challenges to ensure system safety and security, with inherent disturbances to wireless communication networks and also the inevitably larger attack surface that may be exploited by malicious attacks. In this tutorial, experts from academia and industry will introduce the trends and challenges of applying cloud and edge computing for CAVs, highlight representative works in the literature and discuss their limitations, present new promising solutions, and outline future directions in research and engineering. Particular focus will be given to methodologies and tools for building digital twin frameworks with cloud and edge computing for CAVs, quantitative and formal analysis for ensuring CAV safety under disturbances and uncertainties, system-level CAV security threat landscape and defense solution space, and experiences from practical deployment of cloud and edge computing for CAVs.

1

Introduction

Autonomous driving (AD) has seen tremendous progress over the last two decades, propelled by the advancement of neural network-based machine learning techniques and their usage throughout system stages such as perception (Qian *et al.*, 2022), prediction (Jiao *et al.*, 2022; Jiao *et al.*, 2023a; Jiao *et al.*, 2023b), and planning (Liu *et al.*, 2022f; Liu *et al.*, 2023c; Liu *et al.*, 2023a). Significant progress has been made to improve AD performance in various traffic scenarios, including challenging ones such as unprotected left turns, highway merging, and lane changing. Meanwhile, connected vehicle (CV) technologies using vehicular ad-hoc networks enable information sharing among vehicles and surrounding infrastructures (Kenney, 2011; Zheng *et al.*, 2016b). This provides a great complement to the perception and prediction capabilities of individual vehicles, e.g., by sharing out-of-sight information or future motion plans that cannot be accurately predicted.

However, a number of challenges significantly impede the application of AD and CV techniques in realizing connected and automated vehicles (CAVs) in practice. Some of the major challenges include (1) taking full advantage of these techniques to maximize system performance in various applications and scenarios, (2) reducing additional costs to

deploy these techniques, including software and hardware cost, system maintenance and upgrade (Zheng *et al.*, 2016a; Zhu and Sangiovanni-Vincentelli, 2018), (3) ensuring system safety considering environmental uncertainty and disturbance, as well as the difficulty in analyzing neural networks (Zheng *et al.*, 2017a; Zheng *et al.*, 2019; Zhu *et al.*, 2020; Zhu *et al.*, 2021), and (4) defending possible cyber and physical attacks on CAVs (Liu *et al.*, 2023d).

Addressing these challenges and adopting AD and CV techniques require significant computational resources, which could be challenging to deploy on future production vehicles due to the additional cost, energy overhead, hardware maintenance difficulty, etc. The advancement of cloud and edge computing (Varghese *et al.*, 2021; Sánchez *et al.*, 2022; Gross *et al.*, 2022) provides an appealing way to overcome the computational resource limitations of individual vehicles. CAVs can communicate with edge devices that are roadside units (RSUs) and leverage the RSUs' affiliated local sensors and servers, and may also communicate with cloud devices that are far away and can connect to the internet for more information. Tasks with high resource demands and without real-time requirements, e.g., neural network model training, can be uploaded to the cloud, while tasks with moderate resource demands but requiring real-time communication, e.g., perception of the environment and coordination with other vehicles, can be handled by edge computing. Individual vehicles only need to equip the necessary hardware to maintain normal operation and ensure safety when edge and cloud computing are not available occasionally.

Many works have shown that the AD pipelines may perform poorly in long-tailed traffic scenarios (e.g., extreme weather conditions) and are vulnerable to various input noises and attacks such as blemished traffic signs or dirty patches on the road (Chen *et al.*, 2019b; Sato *et al.*, 2021; Zhang *et al.*, 2022). Some methods and frameworks are proposed to enhance the robustness and security of individual modules (Yang *et al.*, 2020; Jiao *et al.*, 2023b; Jiao *et al.*, 2023a) and the AD pipeline (Jiao *et al.*, 2021) of a single autonomous vehicle.

Edge and cloud computing can significantly improve system performance, safety and security by effective collaboration among vehicles and infrastructures. For instance, with the shared information from CAVs,

we can build a Sybil attack detection system (Luo *et al.*, 2021) based on a credibility-enhanced temporal graph convolutional neural network to defend the transportation system against Sybil attacks. At the same time, a cyber-physical credibility framework based on blockchain technology and vehicles' physical sensing capabilities can be maintained to build trust for connected vehicles (Liu *et al.*, 2021h), enabling quick reaction to attacks in a large-scale vehicular network with low resource overhead. Initial studies of such techniques have demonstrated effectiveness in defense against spoofing attacks, bad-mouthing attacks, and Sybil and voting attacks.

System efficiency can be maximized by allocating different tasks to cloud servers, edge devices, or individual vehicles. However, there are some open design challenges. For example, uploading long-tailed training examples to the cloud can build a robust and general model for vehicles in different areas, which can save effort in modeling training, safety and security verification. However, it leads to performance, privacy and security concerns. A general model does not necessarily perform better than a model trained with local traffic data. Individual vehicles may not want to publicize their daily travel trajectories, and even only a small amount of malicious training data can contaminate the model training in the cloud. Another example is collaborative perception with assistance from edge devices. Real-time local maps can be generated on edge devices by collecting shared information from individual vehicles, which can improve system safety, especially in occluded and complex scenarios. However, sharing raw data from Lidars and cameras costs lots of communication resources. Sharing processed information, e.g., detected objects, will lead to challenges in reaching a consensus on these objects and their positions, and each vehicle needs to equip corresponding computational devices.

In the following sections, we will discuss opportunities and challenges for CAVs with cloud and edge computing, as well as present promising solutions to address these challenges and enhance system performance, efficiency, safety and security. In particular, Section 2 presents a mobility digital twin framework at the transportation level for enabling edge and cloud computing for CAVs. Section 3 discusses insights in practical deployment of edge and cloud computing in CAVs. Sections 4 and 5

present software and hardware architecture designs for edge and cloud computing, respectively, that facilitate CAVs. Finally, Section 6 discusses safety and security challenges for cloud- and edge-enabled CAVs and presents potential solutions.

References

- 3GPP. (Accessed: 2022-01-17). “V2X”. URL: <https://www.3gpp.org/v2x>.
- Abbasi, H. I., R. Gholmeh, T. V. Nguyen, S. Patil, and J. Misener. (2022). “LTE-V2X (C-V2X) Performance in Congested Highway Scenarios”. In: *ICC 2022-IEEE International Conference on Communications*. IEEE. 303–308.
- Agarwal, S., N. Snavely, S. M. Seitz, and R. Szeliski. (2010). “Bundle adjustment in the large”. In: *European Conference on Computer Vision*. 29–42.
- Ahmad, H. S., E. Sabouni, W. Xiao, C. G. Cassandras, and W. Li. (2023). “Evaluations of Cyberattacks on Cooperative Control of Connected and Autonomous Vehicles at Bottleneck Points”.
- Arvind, V. Kathail, and K. Pingali. (1980). “A dataflow architecture with tagged tokens”. *Tech. Rep.*
- Bae, S., D. Saxena, A. Nakhaei, C. Choi, K. Fujimura, and S. Moura. (2020). “Cooperation-aware lane change maneuver in dense traffic based on model predictive control with recurrent neural network”. In: *2020 American Control Conference (ACC)*. IEEE. 1209–1216.
- Barendregt, H. P. (1984). *The lambda calculus*. Vol. 3. North-Holland Amsterdam.

- Barrett, R., M. W. Berry, T. F. Chan, J. Demmel, J. Donato, J. Dongarra, V. Eijkhout, R. Pozo, C. Romine, and H. Van der Vorst. (1994). *Templates for the solution of linear systems: building blocks for iterative methods*. Vol. 43. SIAM.
- Cai, Z. and A. Xiong. (2023). “Understand Users’ Privacy Perception and Decision of V2X Communication in Connected Autonomous Vehicles”. In: *Usenix Security Symposium*.
- Cao, Y., N. Wang, C. Xiao, D. Yang, J. Fang, R. Yang, Q. A. Chen, M. Liu, and B. Li. (2021). “Invisible for both Camera and LiDAR: Security of Multi-Sensor Fusion based Perception in Autonomous Driving Under Physical-World Attacks”. In: *IEEE S&P*.
- Cao, Y., C. Xiao, B. Cyr, Y. Zhou, W. Park, S. Rampazzi, Q. A. Chen, K. Fu, and Z. M. Mao. (2019). “Adversarial Sensor Attack on LiDAR-based Perception in Autonomous Driving”. In: *CCS*.
- Cao, Z., E. Biyik, W. Z. Wang, A. Raventos, A. Gaidon, G. Rosman, and D. Sadigh. (2020a). “Reinforcement learning based control of imitative policies for near-accident driving”. *arXiv preprint arXiv:2007.00178*.
- Cao, Z., D. Yang, S. Xu, H. Peng, B. Li, S. Feng, and D. Zhao. (2020b). “Highway exiting planner for automated vehicles using reinforcement learning”. *IEEE Transactions on Intelligent Transportation Systems*.
- Chang, K. K.-C., X. Liu, C.-W. Lin, C. Huang, and Q. Zhu. (2023). “A Safety-Guaranteed Framework for Neural-Network-Based Planners in Connected Vehicles under Communication Disturbance”.
- Chen, J., B. Yuan, and M. Tomizuka. (2019a). “Deep imitation learning for autonomous driving in generic urban scenarios with enhanced safety”. *arXiv preprint arXiv:1903.00640*.
- Chen, L., K. Lu, A. Rajeswaran, K. Lee, A. Grover, M. Laskin, P. Abbeel, A. Srinivas, and I. Mordatch. (2021a). “Decision transformer: Reinforcement learning via sequence modeling”. *Advances in neural information processing systems*. 34: 15084–15097.
- Chen, P.-C., X. Liu, C.-W. Lin, C. Huang, and Q. Zhu. (2023a). “Mixed-Traffic Intersection Management Utilizing Connected and Autonomous Vehicles as Traffic Regulators”. In: *Proceedings of the 28th Asia and South Pacific Design Automation Conference*. 52–57.

- Chen, Q. A., Y. Yin, Y. Feng, Z. M. Mao, and H. X. Liu. (2018). “Exposing Congestion Attack on Emerging Connected Vehicle based Traffic Signal Control”. In: *NDSS*.
- Chen, S.-T., C. Cornelius, J. Martin, and D. H. Chau. (2019b). “Shape-shifter: Robust physical adversarial attack on faster r-cnn object detector”. In: *Machine Learning and Knowledge Discovery in Databases: European Conference, ECML PKDD 2018, Dublin, Ireland, September 10–14, 2018, Proceedings, Part I* 18. Springer. 52–68.
- Chen, S., J. Hu, Y. Shi, L. Zhao, and W. Li. (2020). “A Vision of C-V2X: Technologies, Field Testing, and Challenges With Chinese Development”. *IEEE IoT-J.*
- Chen, X., J. Fan, C. Huang, R. Jiao, W. Li, X. Liu, Y. Wang, Z. Wang, W. Zhou, and Q. Zhu. (2023b). “Safety-Assured Design and Adaptation of Connected and Autonomous Vehicles”.
- Chen, X., C. Fu, F. Zheng, Y. Zhao, H. Li, P. Luo, and G.-J. Qi. (2021b). “A Unified Multi-Scenario Attacking Network for Visual Object Tracking”. In: *AAAI*.
- Cheng, J., G. Yuan, M. Zhou, S. Gao, Z. Huang, and C. Liu. (2020). “A connectivity-prediction-based dynamic clustering model for VANET in an urban scene”. *IEEE Internet of Things Journal*. 7(9): 8410–8418.
- Cheng, M., C. Yin, J. Zhang, S. Nazarian, J. Deshmukh, and P. Bogdan. (2021). “A General Trust Framework for Multi-Agent Systems”. In: *Proceedings of the 20th International Conference on Autonomous Agents and MultiAgent Systems*. 332–340.
- Chiou, H.-P., S. Williams, F. Dellaert, S. Samarasakera, and R. Kumar. (2013). “Robust vision-aided navigation using sliding-window factor graphs”. In: *2013 IEEE International Conference on Robotics and Automation*. IEEE. 46–53.
- Dellaert, F. (2012). “Factor graphs and GTSAM: A hands-on introduction”. *Tech. rep.* Georgia Institute of Technology.
- Dellaert, F. (2021). “Factor graphs: Exploiting structure in robotics”. *Annual Review of Control, Robotics, and Autonomous Systems*. 4: 141–166.
- Dellaert, F. and M. Kaess. (2017). “Factor graphs for robot perception”. *Foundations and Trends® in Robotics*. 6(1-2): 1–139.

- Dennis, J. B. and D. P. Misunas. (1974). “A preliminary architecture for a basic data-flow processor”. In: *Proceedings of the 2nd annual symposium on Computer architecture*. 126–132.
- Dosovitskiy, A., G. Ros, F. Codevilla, A. Lopez, and V. Koltun. (2017). “CARLA: An open urban driving simulator”. In: *Conference on robot learning*. PMLR. 1–16.
- Elbaz, G., T. Avraham, and A. Fischer. (2017). “3D point cloud registration for localization using a deep neural network auto-encoder”. In: *CVPR*. 4631–4640.
- Fan, J., C. Huang, X. Chen, W. Li, and Q. Zhu. (2020). “ReachNN*: A Tool for Reachability Analysis of Neural-Network Controlled Systems”. In: *Automated Technology for Verification and Analysis*. Ed. by D. V. Hung and O. Sokolsky. Cham: Springer International Publishing. 537–542.
- Fang, W., Y. Zhang, B. Yu, and S. Liu. (2017). “FPGA-based ORB feature extraction for real-time visual SLAM”. In: *ICFPT*. IEEE. 275–278.
- Fang, W., Y. Zhang, B. Yu, and S. Liu. (2018). “DragonFly+: FPGA-based quad-camera visual SLAM system for autonomous vehicles”. *Proc. IEEE HotChips*. 1.
- Ford. (2021). *Adaptive Cruise Control*. URL: <https://www.ford.com/technology/driver-assist-technology/adaptive-cruise-control/> (accessed on 09/12/2021).
- Forster, C., L. Carlone, F. Dellaert, and D. Scaramuzza. (2016). “On-Manifold Preintegration for Real-Time Visual–Inertial Odometry”. *IEEE Trans. Robot.* 33(1): 1–21.
- Gan, Y., Y. Bo, B. Tian, L. Xu, W. Hu, S. Liu, Q. Liu, Y. Zhang, J. Tang, and Y. Zhu. (2021). “Eudoxus: Characterizing and accelerating localization in autonomous machines industry track paper”. In: *2021 IEEE International Symposium on High-Performance Computer Architecture (HPCA)*. IEEE. 827–840.
- Gan, Y., P. Whatmough, J. Leng, B. Yu, S. Liu, and Y. Zhu. (2022). “Braum: Analyzing and protecting autonomous machine software stack”. In: *2022 IEEE 33rd International Symposium on Software Reliability Engineering (ISSRE)*. IEEE. 85–96.

- Gao, T., Z. Wan, Y. Zhang, B. Yu, Y. Zhang, S. Liu, and A. Raychowdhury. (2021). “IELAS: An ELAS-based energy-efficient accelerator for real-time stereo matching on FPGA platform”. In: *2021 IEEE 3rd International Conference on Artificial Intelligence Circuits and Systems (AICAS)*. IEEE. 1–4.
- Geng, T., M. Amaris, S. Zuckerman, A. Goldman, G. R. Gao, and J.-L. Gaudiot. (2022). “A Profile-Based AI-Assisted Dynamic Scheduling Approach for Heterogeneous Architectures”. *International Journal of Parallel Programming*. 50(1): 115–151.
- Grand View Research. (Accessed: 2021-12-09). “Digital Twin Market Size, Share & Trends Analysis Report By End-use (Automotive & Transport, Retail & Consumer Goods, Agriculture, Manufacturing, Energy & Utilities), By Region, And Segment Forecasts, 2021 - 2028”. URL: <https://www.grandviewresearch.com/industry-analysis/digital-twin-market>.
- Griewank, A. (1989). “On automatic differentiation”. *Mathematical Programming: recent developments and applications*. 6(6): 83–107.
- Gross, J., M. Törngren, G. Dán, D. Broman, E. Herzog, I. Leite, R. Ramakrishna, R. Stower, and H. Thompson. (2022). “TECoSA – Trends, Drivers, and Strategic Directions for Trustworthy Edge Computing in Industrial Applications”. *INSIGHT*. 25(4): 29–34. DOI: [10.1002/inst.12408](https://doi.org/10.1002/inst.12408).
- Guan, H., S. Liu, X. Ma, W. Niu, B. Ren, X. Shen, Y. Wang, and P. Zhao. (2021). “CoCoPIE: Enabling real-time AI on off-the-shelf mobile devices via compression-compilation co-design”. *Communications of the ACM*. 64(6): 62–68.
- Guo, Y., X. Hu, B. Hu, J. Cheng, M. Zhou, and R. Y. Kwok. (2017). “Mobile cyber physical systems: Current challenges and future networking applications”. *IEEE Access*. 6: 12360–12368.
- Gupta, A., L. Fan, S. Ganguli, and L. Fei-Fei. (2022). “Metamorph: Learning universal controllers with transformers”. *arXiv preprint arXiv:2203.11931*.
- Hao, Y., Y. Gan, B. Yu, Q. Liu, S. Liu, and Y. Zhu. (2023). “BLITZ-CRANK: Factor Graph Accelerator for Motion Planning”. In: *2023 60th ACM/IEEE Design Automation Conference (DAC)*. IEEE.

- Hao, Y., B. Yu, Q. Liu, S. Liu, and Y. Zhu. (2022). “Factor Graph Accelerator for LiDAR-Inertial Odometry”. In: *Proceedings of the 41st IEEE/ACM International Conference on Computer-Aided Design*. 1–7.
- Higuchi, T., S. Ucar, and O. Altintas. (2019). “A Collaborative Approach to Finding Available Parking Spots”. In: *2019 IEEE 90th Vehicular Technology Conference (VTC2019-Fall)*. IEEE. 1–5.
- Hu, S., Q. A. Chen, J. Sun, Y. Feng, Z. M. Mao, and H. X. Liu. (2021). “Automated Discovery of Denial-of-Service Vulnerabilities in Connected Vehicle Protocols”. In: *Usenix Security*.
- Huang, C., K.-C. Chang, C.-W. Lin, and Q. Zhu. (2020a). “SAW: A Tool for Safety Analysis of Weakly-hard Systems”. *32nd International Conference on Computer-Aided Verification (CAV’20)*.
- Huang, C., J. Fan, X. Chen, W. Li, and Q. Zhu. (2020b). “Divide and Slide: Layer-Wise Refinement for Output Range Analysis of Deep Neural Networks”. *IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems*. 39(11): 3323–3335. doi: [10.1109/TCAD.2020.3013071](https://doi.org/10.1109/TCAD.2020.3013071).
- Huang, C., J. Fan, X. Chen, W. Li, and Q. Zhu. (2022a). “POLAR: A Polynomial Arithmetic Framework for Verifying Neural-Network Controlled Systems”. In: *Automated Technology for Verification and Analysis*. Springer International Publishing.
- Huang, C., J. Fan, W. Li, X. Chen, and Q. Zhu. (2019a). “ReachNN: Reachability analysis of neural-network controlled systems”. *ACM Transactions on Embedded Computing Systems (TECS)*. 18(5s): 1–22.
- Huang, C., W. Li, and Q. Zhu. (2019b). “Formal Verification of Weakly-Hard Systems”. In: *the 22nd ACM International Conference on Hybrid Systems: Computation and Control (HSCC)*.
- Huang, C., K. Wardega, W. Li, and Q. Zhu. (2019c). “Exploring weakly-hard paradigm for networked systems”. In: *Proceedings of the Workshop on Design Automation for CPS and IoT*. 51–59.
- Huang, C., S. Xu, Z. Wang, S. Lan, W. Li, and Q. Zhu. (2020c). “Opportunistic Intermittent Control with Safety Guarantees for Autonomous Systems”. *Design Automation Conference (DAC’20)*.

- Huang, P.-Y., K.-W. Liu, Z.-L. Li, S. Park, E. Andert, C.-W. Lin, and A. Shrivastava. (2022b). “Compatibility checking for autonomous lane-changing assistance systems”. In: *2022 Design, Automation & Test in Europe Conference & Exhibition (DATE)*. IEEE. 1161–1164.
- Huang, Y., W. Zheng, Y. Zhang, J. Zhou, and J. Lu. (2023). “Tri-perspective view for vision-based 3d semantic occupancy prediction”. In: *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition*. 9223–9232.
- Indelman, V., S. Williams, M. Kaess, and F. Dellaert. (2012). “Factor graph based incremental smoothing in inertial navigation systems”. In: *2012 15th International Conference on Information Fusion*. IEEE. 2154–2161.
- Janner, M., Q. Li, and S. Levine. (2021). “Offline reinforcement learning as one big sequence modeling problem”. *Advances in neural information processing systems*. 34: 1273–1286.
- Jia, W., Z. Lu, H. Zhang, Z. Liu, J. Wang, and G. Qu. (2022). “Fooling the Eyes of Autonomous Vehicles: Robust Physical Adversarial Examples Against Traffic Sign Recognition Systems”. In: *NDSS*.
- Jia, Y. J., Y. Lu, J. Shen, Q. A. Chen, H. Chen, Z. Zhong, and T. W. Wei. (2020). “Fooling Detection Alone is Not Enough: Adversarial Attack against Multiple Object Tracking”. In: *ICLR*.
- Jiang, X., X. Luo, N. Guan, Z. Dong, S. Liu, and W. Yi. (2023). “Analysis and Optimization of Worst-Case Time Disparity in Cause-Effect Chains”. In: *2023 Design, Automation & Test in Europe Conference & Exhibition (DATE)*. IEEE. 1–6.
- Jiao, J. (2018). “Machine learning assisted high-definition map creation”. In: *2018 IEEE 42nd Annual Computer Software and Applications Conference (COMPSAC)*. Vol. 1. IEEE. 367–373.
- Jiao, R., J. Bai, X. Liu, T. Sato, X. Yuan, Q. A. Chen, and Q. Zhu. (2023a). “Learning Representation for Anomaly Detection of Vehicle Trajectories”. In: *2023 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*. IEEE.
- Jiao, R., H. Liang, T. Sato, J. Shen, Q. A. Chen, and Q. Zhu. (2021). “End-to-end uncertainty-based mitigation of adversarial attacks to automated lane centering”. In: *2021 IEEE Intelligent Vehicles Symposium (IV)*. IEEE. 266–273.

- Jiao, R., X. Liu, T. Sato, Q. A. Chen, and Q. Zhu. (2023b). “Semi-supervised Semantics-guided Adversarial Training for Trajectory Prediction”. In: *2023 IEEE/CVF International Conference on Computer Vision (ICCV)*. IEEE.
- Jiao, R., X. Liu, B. Zheng, D. Liang, and Q. Zhu. (2022). “Tae: A semi-supervised controllable behavior-aware trajectory generator and predictor”. In: *2022 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*. IEEE. 12534–12541.
- Joseph, M. and P. Pandya. (1986). “Finding response times in a real-time system”. *The Computer Journal*. 29(5): 390–395.
- Kaess, M., H. Johannsson, R. Roberts, V. Ila, J. J. Leonard, and F. Dellaert. (2012). “iSAM2: Incremental smoothing and mapping using the Bayes tree”. *The International Journal of Robotics Research*. 31(2): 216–235.
- Kaess, M., A. Ranganathan, and F. Dellaert. (2008). “iSAM: Incremental Smoothing and Mapping”. *IEEE Transactions on Robotics*. 24: 1365–1378.
- Kenney, J. B. (2011). “Dedicated Short-Range Communications (DSRC) Standards in the United States”. *Proceedings of the IEEE*. 99(7): 1162–1182. DOI: [10.1109/JPROC.2011.2132790](https://doi.org/10.1109/JPROC.2011.2132790).
- Koopman, P. (2022). *How Safe Is Safe Enough?: Measuring and Predicting Autonomous Vehicle Safety*.
- Kumar, S. A., R. Madhumathi, P. R. Chelliah, L. Tao, and S. Wang. (2018). “A novel digital twin-centric approach for driver intention prediction and traffic congestion avoidance”. *Journal of Reliable Intelligent Environments*. 4(4): 199–209.
- Kümmerle, R., G. Grisetti, H. Strasdat, K. Konolige, and W. Burgard. (2011). “g 2 o: A general framework for graph optimization”. In: *2011 IEEE International Conference on Robotics and Automation*. IEEE. 3607–3613.
- Lattner, C., M. Amini, U. Bondhugula, A. Cohen, A. Davis, J. Pienaar, R. Riddle, T. Shpeisman, N. Vasilache, and O. Zinenko. (2020). “MLIR: A compiler infrastructure for the end of Moore’s law”. *arXiv preprint arXiv:2002.11054*.

- Lattner, C., M. Amini, U. Bondhugula, A. Cohen, A. Davis, J. Pienaar, R. Riddle, T. Shpeisman, N. Vasilache, and O. Zinenko. (2021). “Mlir: Scaling compiler infrastructure for domain specific computation”. In: *2021 IEEE/ACM International Symposium on Code Generation and Optimization (CGO)*. IEEE. 2–14.
- Li, B., S. Liu, J. Tang, J.-L. Gaudiot, L. Zhang, and Q. Kong. (2020). “Autonomous last-mile delivery vehicles in complex traffic environments”. *Computer*. 53(11): 26–35.
- Li, D., J. Tang, and S. Liu. (2021a). “Brief Industry Paper: An Edge-based High-Definition Map Crowdsourcing Task Distribution Framework for Autonomous Driving”. In: *2021 IEEE 27th Real-Time and Embedded Technology and Applications Symposium (RTAS)*. IEEE. 453–456.
- Li, J., L. Sun, M. Tomizuka, and W. Zhan. (2021b). “A Safe Hierarchical Planning Framework for Complex Driving Scenarios based on Reinforcement Learning”. *arXiv preprint arXiv:2101.06778*.
- Li, Q., Y. Wang, Y. Wang, and H. Zhao. (2022a). “Hdmapnet: An online hd map construction and evaluation framework”. In: *2022 International Conference on Robotics and Automation (ICRA)*. IEEE. 4628–4634.
- Li, Y., Z. Ge, G. Yu, J. Yang, Z. Wang, Y. Shi, J. Sun, and Z. Li. (2022b). “Bevdepth: Acquisition of reliable depth for multi-view 3d object detection”. *arXiv preprint arXiv:2206.10092*.
- Li, Z., W. Wang, H. Li, E. Xie, C. Sima, T. Lu, Y. Qiao, and J. Dai. (2022c). “Bevformer: Learning bird’s-eye-view representation from multi-camera images via spatiotemporal transformers”. In: *Computer Vision–ECCV 2022: 17th European Conference, Tel Aviv, Israel, October 23–27, 2022, Proceedings, Part IX*. Springer. 1–18.
- Liang, H., Z. Wang, R. Jiao, and Q. Zhu. (2020). “Leveraging Weakly-hard Constraints for Improving System Fault Tolerance with Functional and Timing Guarantees”. In: *2020 IEEE/ACM International Conference On Computer Aided Design (ICCAD)*. 1–9.
- Liang, H., Z. Wang, D. Roy, S. Dey, S. Chakraborty, and Q. Zhu. (2019). “Security-driven Codesign with Weakly-hard Constraints for Real-time Embedded Systems”. In: *37th IEEE International Conference on Computer Design (ICCD’19)*.

- Liao, X., Z. Wang, X. Zhao, Z. Zhao, K. Han, P. Tiwari, M. Barth, and G. Wu. (2022). “Online Prediction of Lane Change with a Hierarchical Learning-Based Approach”. In: *Proceedings 2022 IEEE International Conference on Robotics and Automation*. IEEE.
- Liu, C. L. and J. W. Layland. (1973). “Scheduling algorithms for multiprogramming in a hard-real-time environment”. *Journal of the ACM (JACM)*. 20(1): 46–61.
- Liu, L., J. Tang, S. Liu, B. Yu, Y. Xie, and J.-L. Gaudiot. (2021a). “ π -rt: A runtime framework to enable energy-efficient real-time robotic vision applications on heterogeneous architectures”. *Computer*. 54(4): 14–25.
- Liu, Q., S. Qin, B. Yu, J. Tang, and S. Liu. (2020a). “ π -BA: Bundle Adjustment Hardware Accelerator Based on Distribution of 3D-Point Observations”. *IEEE Transactions on Computers*. 69(7): 1083–1095.
- Liu, Q., Z. Wan, B. Yu, W. Liu, S. Liu, and A. Raychowdhury. (2022a). “An energy-efficient and runtime-reconfigurable fpga-based accelerator for robotic localization systems”. In: *2022 IEEE Custom Integrated Circuits Conference (CICC)*. IEEE. 01–02.
- Liu, S. (2020a). “Critical business decision making for technology startups: A perceptin case study”. *IEEE Engineering Management Review*. 48(4): 32–36.
- Liu, S. (2020b). *Engineering autonomous vehicles and robots: the drag-onfly modular-based approach*. John Wiley & Sons.
- Liu, S. and J.-L. Gaudiot. (2020). “Autonomous vehicles lite self-driving technologies should start small, go slow”. *IEEE Spectrum*. 57(3): 36–49.
- Liu, S. and J.-L. Gaudiot. (2022). “Rise of the autonomous machines”. *Computer*. 55(1): 64–73.
- Liu, S., J.-L. Gaudiot, and H. Kasahara. (2021b). “Engineering education in the age of autonomous machines”. *Computer*. 54(4): 66–69.
- Liu, S., Y. Huang, A. Kong, J. Tang, and X. Liu. (2022b). “Rise of the Automotive Health-Domain Controllers: Empowering Healthcare Services in Intelligent Vehicles”. *IEEE Internet of Things Journal*. 9(24): 24882–24889.

- Liu, S., Y. Huang, and L. Shi. (2022c). “Autonomous Mobile Clinics: Empowering Affordable Anywhere Anytime Healthcare Access”. *IEEE Engineering Management Review*.
- Liu, S., A. Kong, Y. Huang, and X. Liu. (2022d). “Autonomous mobile clinics”. *Bulletin of the World Health Organization*. 100(9): 527–527A.
- Liu, S., L. Li, J. Tang, S. Wu, and J.-L. Gaudiot. (2020b). “Creating autonomous vehicle systems”. *Synthesis Lectures on Computer Science*. 8(2): i–216.
- Liu, S., X. Li, T. Geng, S. Zuckerman, and J.-L. Gaudiot. (2022e). “Programming Autonomous Machines: Special Session Paper”. In: *2022 International Conference on Embedded Software (EMSOFT)*. IEEE. 24–33.
- Liu, S., L. Liu, J. Tang, B. Yu, Y. Wang, and W. Shi. (2019). “Edge computing for autonomous driving: Opportunities and challenges”. *Proceedings of the IEEE*. 107(8): 1697–1716.
- Liu, S., J. Peng, and J.-L. Gaudiot. (2017a). “Computer, drive my car!” *Computer*. 50(01): 8–8.
- Liu, S., R. N. Pittman, A. Forin, and J.-L. Gaudiot. (2013). “Achieving energy efficiency through runtime partial reconfiguration on reconfigurable systems”. *ACM Transactions on Embedded Computing Systems (TECS)*. 12(3): 1–21.
- Liu, S., J. Tang, C. Wang, Q. Wang, and J.-L. Gaudiot. (2017b). “A unified cloud platform for autonomous driving”. *Computer*. 50(12): 42–49.
- Liu, S., J. Tang, Z. Zhang, and J.-L. Gaudiot. (2017c). “Computer architectures for autonomous driving”. *Computer*. 50(8): 18–25.
- Liu, S., Z. Wan, B. Yu, and Y. Wang. (2021c). “Robotic computing on fpgas”. *Synthesis Lectures on Computer Architecture*. 16(1): 1–218.
- Liu, S., B. Yu, N. Guan, Z. Dong, and B. Akesson. (2021d). “Real-Time Scheduling and Analysis of an Autonomous Driving System”. In: *42nd IEEE Real-Time Systems Symposium (RTSS) Industry Challenge*. IEEE.

- Liu, S., B. Yu, Y. Liu, K. Zhang, Y. Qiao, T. Y. Li, J. Tang, and Y. Zhu. (2021e). “Brief industry paper: The matter of time—A general and efficient system for precise sensor synchronization in robotic computing”. In: *2021 IEEE 27th Real-Time and Embedded Technology and Applications Symposium (RTAS)*. IEEE. 413–416.
- Liu, S., Y. Zhu, B. Yu, J.-L. Gaudiot, and G. R. Gao. (2021f). “Dataflow accelerator architecture for autonomous machine computing”. *arXiv preprint arXiv:2109.07047*.
- Liu, W., B. Yu, Y. Gan, Q. Liu, J. Tang, S. Liu, and Y. Zhu. (2021g). “Archytas: A framework for synthesizing and dynamically optimizing accelerators for robotic localization”. In: *MICRO-54: 54th Annual IEEE/ACM International Symposium on Microarchitecture*. 479–493.
- Liu, X., C. Huang, Y. Wang, B. Zheng, and Q. Zhu. (2022f). “Physics-aware safety-assured design of hierarchical neural network based planner”. In: *2022 ACM/IEEE 13th International Conference on Cyber-Physical Systems (ICCPs)*. IEEE. 137–146.
- Liu, X., R. Jiao, Y. Wang, Y. Han, B. Zheng, and Q. Zhu. (2023a). “Safety-Assured Speculative Planning with Adaptive Prediction”. In: *2023 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*. IEEE.
- Liu, X., R. Jiao, B. Zheng, D. Liang, and Q. Zhu. (2023b). “Connectivity Enhanced Safe Neural Network Planner for Lane Changing in Mixed Traffic”. *arXiv preprint arXiv:2302.02513*.
- Liu, X., R. Jiao, B. Zheng, D. Liang, and Q. Zhu. (2023c). “Safety-Driven Interactive Planning for Neural Network-Based Lane Changing”. In: *Proceedings of the 28th Asia and South Pacific Design Automation Conference. ASPDAC '23*. Tokyo, Japan. DOI: [10.1145/3566097.3567847](https://doi.org/10.1145/3566097.3567847).
- Liu, X., B. Luo, A. Abdo, N. Abu-Ghazaleh, and Q. Zhu. (2021h). “Securing connected vehicle applications with an efficient dual cyber-physical blockchain framework”. In: *2021 IEEE Intelligent Vehicles Symposium (IV)*. IEEE. 393–400.

- Liu, X., Y. Luo, A. Goeckner, T. Chakraborty, R. Jiao, N. Wang, Y. Wang, T. Sato, Q. A. Chen, and Q. Zhu. (2023d). “Invited: Waving the Double-Edged Sword: Building Resilient CAVs with Edge and Cloud Computing”. In: *Proceedings of the 60th Annual Design Automation Conference*. 1–4.
- Liu, X., N. Masoud, and Q. Zhu. (2020c). “Impact of sharing driving attitude information: A quantitative study on lane changing”. In: *2020 IEEE Intelligent Vehicles Symposium (IV)*. IEEE. 1998–2005.
- Liu, X., N. Masoud, Q. Zhu, and A. Khojandi. (2022g). “A markov decision process framework to incorporate network-level data in motion planning for connected and automated vehicles”. *Transportation Research Part C: Emerging Technologies*. 136: 103550.
- Liu, X., G. Zhao, N. Masoud, and Q. Zhu. (2021i). “Trajectory Planning for Connected and Automated Vehicles: Cruising, Lane Changing, and Platooning”. *SAE International Journal of Connected and Automated Vehicles*. 4(12-04-04-0025): 315–333.
- Liu, Y., J. Zhang, L. Fang, Q. Jiang, and B. Zhou. (2021j). “Multi-modal motion prediction with stacked transformers”. In: *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition*. 7577–7586.
- Lourakis, M. and A. A. Argyros. (2005). “Is Levenberg-Marquardt the most efficient optimization algorithm for implementing bundle adjustment?” In: *Computer Vision, 2005. ICCV 2005. Tenth IEEE International Conference on*. Vol. 2. IEEE. 1526–1531.
- Lowe, D. G. (1999). “Object recognition from local scale-invariant features”. In: *ICCV*. Vol. 2. Ieee. 1150–1157.
- Luo, B., X. Liu, and Q. Zhu. (2021). “Credibility enhanced temporal graph convolutional network based sybil attack detection on edge computing servers”. In: *2021 IEEE Intelligent Vehicles Symposium (IV)*. IEEE. 524–531.
- Luo, M., A. C. Myers, and G. E. Suh. (2020). “Stealthy Tracking of Autonomous Vehicles with Cache Side Channels”. In: *USENIX Security*.

- Luo, Y., N. Wang, B. Yu, S. Liu, and Q. A. Chen. (2022). "Infrastructure-Aided Defense for Autonomous Driving Systems: Opportunities and Challenges". In: *NDSS Workshop on Automotive and Autonomous Vehicle Security (AutoSec)*.
- Ma, H., J. Chen, S. E. Li, Z. Lin, Y. Guan, Y. Ren, and S. Zheng. (2021). "Model-based Constrained Reinforcement Learning using Generalized Control Barrier Function". *arXiv preprint arXiv:2103.01556*.
- Ma, Y., Z. Wang, H. Yang, and L. Yang. (2020). "Artificial intelligence applications in the development of autonomous vehicles: a survey". *IEEE/CAA Journal of Automatica Sinica*. 7(2): 315–329.
- Man, Y., M. Li, and R. Gerdes. (2020). "GhostImage: Remote Perception Attacks against Camera-based Image Classification Systems". In: *RAID*.
- Markolf, L., J. Eilbrecht, and O. Stursberg. (2020). "Trajectory Planning for Autonomous Vehicles combining Nonlinear Optimal Control and Supervised Learning". *IFAC-PapersOnLine*. 53(2): 15608–15614.
- Masuda, T., B. Kim, and S. Shiraishi. (2019). "Proactive vehicle maintenance scheduling based on digital twin simulations".
- Mirchevska, B., C. Pek, M. Werling, M. Althoff, and J. Boedecker. (2018). "High-level decision making for safe and reasonable autonomous lane changing using reinforcement learning". In: *2018 21st International Conference on Intelligent Transportation Systems (ITSC)*. IEEE. 2156–2162.
- Mur-Artal, R. and J. D. Tardós. (2017). "Orb-slam2: An open-source slam system for monocular, stereo, and rgb-d cameras". *IEEE Trans. Robot.* 33(5): 1255–1262.
- Naveed, K. B., Z. Qiao, and J. M. Dolan. (2020). "Trajectory Planning for Autonomous Vehicles Using Hierarchical Reinforcement Learning". *arXiv preprint arXiv:2011.04752*.
- NHTSA. (2022). "Summary Report: Standing General Order on Crash Reporting for Automated Driving Systems".
- Nosrati, M. S., E. A. Abolfathi, M. Elmahgiubi, P. Yadmallat, J. Luo, Y. Zhang, H. Yao, H. Zhang, and A. Jamil. (2018). "Towards practical hierarchical reinforcement learning for multi-lane autonomous driving".

- Orzechowski, P. F., A. Meyer, and M. Lauer. (2018). “Tackling occlusions & limited sensor range with set-based safety verification”. In: *2018 21st International Conference on Intelligent Transportation Systems (ITSC)*. IEEE. 1729–1736.
- Palacharla, S., N. P. Jouppi, and J. E. Smith. (1997). “Complexity-effective superscalar processors”. In: *Proceedings of the 24th annual international symposium on Computer architecture*. 206–218.
- Pan, Y., N. Wu, T. Qu, P. Li, K. Zhang, and H. Guo. (2020). “Digital twin-driven production logistics synchronization system for vehicle routing problems with pick-up and delivery in industrial park”. *International Journal of Computer Integrated Manufacturing*: 1–15.
- Papageorgiou, M., E. Kosmatopoulos, and I. Papamichail. (2008). “Effects of variable speed limits on motorway traffic flow”. *Transportation Research Record*. 2047(1): 37–48.
- Patt, Y. N., W.-m. Hwu, and M. Shebanow. (1985). “HPS, a new microarchitecture: Rationale and introduction”. *ACM SIGMICRO Newsletter*. 16(4): 103–108.
- PerceptIn. (2021). “RTSS 2021 Industry Challenge”. URL: <http://2021.rtss.org/wp-content/uploads/2021/06/RTSS2021-Industry-Challenge-v2.pdf>.
- Qian, R., X. Lai, and X. Li. (2022). “3d object detection for autonomous driving: a survey”. *Pattern Recognition*. 130: 108796.
- Qin, S., Q. Liu, B. Yu, and S. Liu. (2019). “ π -BA: Bundle Adjustment Acceleration on Embedded FPGAs with Co-observation Optimization”. In: *2019 IEEE 27th Annual International Symposium on Field-Programmable Custom Computing Machines (FCCM)*. IEEE. 100–108.
- Quigley, M., K. Conley, B. Gerkey, J. Faust, T. Foote, J. Leibs, R. Wheeler, and A. Y. Ng. (2009). “ROS: an open-source Robot Operating System”. In: *ICRA workshop on open source software*. Vol. 3. No. 3.2. Kobe, Japan. 5.
- Sánchez, J. M. G., N. Jörgensen, M. Törngren, R. Inam, A. Berezovskyi, L. Feng, E. Fersman, M. R. Ramli, and K. Tan. (2022). “Edge Computing for Cyber-Physical Systems: A Systematic Mapping Study Emphasizing Trustworthiness”. *ACM Trans. Cyber-Phys. Syst.* 6(3). DOI: [10.1145/3539662](https://doi.org/10.1145/3539662).

- Santa, J., J. Ortiz, P. J. Fernandez, M. Luis, C. Gomes, J. Oliveira, D. Gomes, R. Sanchez-Iborra, S. Sargent, and A. F. Skarmeta. (2020). “MIGRATE: Mobile Device Virtualisation Through State Transfer”. *IEEE Access*. 8: 25848–25862. DOI: [10.1109/ACCESS.2020.2971090](https://doi.org/10.1109/ACCESS.2020.2971090).
- Santa, J., P. J. Fernández, J. Ortiz, R. Sanchez-Iborra, and A. F. Skarmeta. (2019). “SURROGATES: Virtual OBUs to Foster 5G Vehicular Services”. *Electronics*. 8(2). DOI: [10.3390/electronics8020117](https://doi.org/10.3390/electronics8020117).
- Sato, T., J. Shen, N. Wang, Y. Jia, X. Lin, and Q. A. Chen. (2021). “Dirty road can attack: Security of deep learning based automated lane centering under physical-world attack”. In: *Proceedings of the 30th USENIX Security Symposium (USENIX Security 21)*.
- Shan, T., B. Englot, D. Meyers, W. Wang, C. Ratti, and D. Rus. (2020). “Lio-sam: Tightly-coupled lidar inertial odometry via smoothing and mapping”. In: *2020 IEEE/RSJ international conference on intelligent robots and systems (IROS)*. IEEE. 5135–5142.
- Shen, J., J. Y. Won, Z. Chen, and Q. A. Chen. (2020). “Drift with Devil: Security of Multi-Sensor Fusion based Localization in High-Level Autonomous Driving under GPS Spoofing”. In: *USENIX Security*.
- Shou, Z., Z. Wang, K. Han, Y. Liu, P. Tiwari, and X. Di. (2020). “Long-Term Prediction of Lane Change Maneuver Through a Multilayer Perceptron”. In: *IEEE Intelligent Vehicles Symposium (IV)*.
- Silva, L., N. Magaia, B. Sousa, A. Kobusińska, A. Casimiro, C. X. Mavromoustakis, G. Mastorakis, and V. H. C. De Albuquerque. (2021). “Computing Paradigms in Emerging Vehicular Environments: A Review”. *IEEE/CAA Journal of Automatica Sinica*. 8(3): 491–511.
- Stein, W. J. and T. R. Neuman. (2017). “Mitigation Strategies for Design Exceptions”. *US Federal Highway Administration. Office of Safety*.
- Strasdat, H., J. Montiel, and A. J. Davison. (2010). “Real-time monocular SLAM: Why filter?” In: *Robotics and Automation (ICRA), 2010 IEEE International Conference on*. IEEE. 2657–2664.
- Sun, J., Y. Cao, Q. A. Chen, and Z. M. Mao. (2020). “Towards Robust LiDAR-based Perception in Autonomous Driving: General Black-box Adversarial Sensor Attack and Countermeasures”. In: *USENIX Security*.

- Sun, L., C. Tang, Y. Niu, E. Sachdeva, C. Cho, T. Misu, M. Tomizuka, and W. Zhan. (2022). “Domain Knowledge Driven Pseudo Labels for Interpretable Goal-Conditioned Interactive Trajectory Prediction”. *arXiv preprint arXiv:2203.15112*.
- Sun, Z., S. Balakrishnan, L. Su, A. Bhuyan, P. Wang, and C. Qiao. (2021). “Who Is in Control? Practical Physical Layer Attack and Defense for mmWave-Based Sensing in Autonomous Vehicles”. *IEEE TIFS*.
- Sung, H.-H., Y. Xu, J. Guan, W. Niu, B. Ren, Y. Wang, S. Liu, and X. Shen. (2022). “Brief industry paper: Enabling level-4 autonomous driving on a single \$1 k off-the-shelf card”. In: *2022 IEEE 28th Real-Time and Embedded Technology and Applications Symposium (RTAS)*. IEEE. 297–300.
- Tang, J., S. Liu, L. Liu, B. Yu, and W. Shi. (2020a). “LoPECS: A low-power edge computing system for real-time autonomous driving services”. *IEEE Access*. 8: 30467–30479.
- Tang, J., B. Yu, S. Liu, Z. Zhang, W. Fang, and Y. Zhang. (2018). “ π -soc: Heterogeneous soc architecture for visual inertial slam applications”. In: *2018 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*. IEEE. 8302–8307.
- Tang, J., R. Yu, S. Liu, and J.-L. Gaudiot. (2020b). “A container based edge offloading framework for autonomous driving”. *ieee Access*. 8: 33713–33726.
- Tang, J., W. Zhu, X. Li, S. Liu, and X. Liu. (2022). “A Distributed Pricing Strategy for Edge Computation Offloading Optimization in Autonomous Driving”. *IEEE Network*.
- Tang, K., J. Shen, and Q. A. Chen. (2021). “Fooling Perception via Location: A Case of Region-of-Interest Attacks on Traffic Light Detection in Autonomous Driving”. In: *AutoSec*.
- Topcuoglu, H., S. Hariri, and M.-Y. Wu. (2002). “Performance-effective and low-complexity task scheduling for heterogeneous computing”. *IEEE Transactions on Parallel and Distributed Systems*. 13(3): 260–274. DOI: [10.1109/71.993206](https://doi.org/10.1109/71.993206).
- Toyota. (2021). *Toyota Safety Sense: The Standard for Safety*. URL: <https://www.toyota.com/safety-sense/> (accessed on 09/12/2021).

- Triggs, B., P. F. McLauchlan, R. I. Hartley, and A. W. Fitzgibbon. (1999). *Bundle adjustment—a modern synthesis*. Springer Berlin Heidelberg. 298–372.
- Tripathy, A. K., P. K. Tripathy, A. G. Mohapatra, N. K. Ray, and S. P. Mohanty. (2020). “WeDoShare: A Ridesharing Framework in Transportation Cyber-Physical System for Sustainable Mobility in Smart Cities”. *IEEE Consumer Electronics Magazine*. 9(4): 41–48.
- Tyagi, A., Y. Gan, S. Liu, B. Yu, P. Whatmough, and Y. Zhu. (2023). “Thales: Formulating and Estimating Architectural Vulnerability Factors for DNN Accelerators”. In: *2023 IEEE International Symposium on High-Performance Computer Architecture (HPCA)*. IEEE.
- Varghese, B., E. de Lara, A. Y. Ding, C.-H. Hong, F. Bonomi, S. Dustdar, P. Harvey, P. Hewkin, W. Shi, M. Thiele, and P. Willis. (2021). “Revisiting the Arguments for Edge Computing Research”. *IEEE Internet Computing*. 25(5): 36–42. doi: [10.1109/MIC.2021.3093924](https://doi.org/10.1109/MIC.2021.3093924).
- Volkswagen. (2021). *Adaptive Cruise Control*. URL: <https://www.volkswagen-newsroom.com/en/adaptive-cruise-control-acc-3664> (accessed on 09/12/2021).
- Wan, Z., A. Lele, B. Yu, S. Liu, Y. Wang, V. J. Reddi, C. Hao, and A. Raychowdhury. (2022a). “Robotic computing on fpgas: Current progress, research challenges, and opportunities”. In: *2022 IEEE 4th International Conference on Artificial Intelligence Circuits and Systems (AICAS)*. IEEE. 291–295.
- Wan, Z., B. Yu, T. Y. Li, J. Tang, Y. Zhu, Y. Wang, A. Raychowdhury, and S. Liu. (2021a). “A survey of fpga-based robotic computing”. *IEEE Circuits and Systems Magazine*. 21(2): 48–74.
- Wan, Z., Y. Zhang, A. Raychowdhury, B. Yu, Y. Zhang, and S. Liu. (2021b). “An energy-efficient quad-camera visual system for autonomous machines on fpga platform”. In: *2021 IEEE 3rd International Conference on Artificial Intelligence Circuits and Systems (AICAS)*. IEEE. 1–4.
- Wan, Z., J. Shen, J. Chuang, X. Xia, J. Garcia, J. Ma, and Q. A. Chen. (2022b). “Too Afraid to Drive: Systematic Discovery of Denial-of-Service Vulnerability in Autonomous Driving Planning under Physical-World Attacks”. In: *NDSS*.

- Wang, D., C. Li, S. Wen, Q.-L. Han, S. Nepal, X. Zhang, and Y. Xiang. (2021a). “Daedalus: Breaking nonmaximum suppression in object detection via adversarial examples”. *IEEE Trans. Cybern.*
- Wang, H., T. Liu, B. Kim, C.-W. Lin, S. Shiraishi, J. Xie, and Z. Han. (2020a). “Architectural design alternatives based on cloud/edge/fog computing for connected vehicles”. *IEEE Communications Surveys & Tutorials*. 22(4): 2349–2377.
- Wang, J., Y. Wang, D. Zhang, Y. Yang, and R. Xiong. (2020b). “Learning hierarchical behavior and motion planning for autonomous driving”. *arXiv preprint arXiv:2005.03863*.
- Wang, P., C.-Y. Chan, and A. de La Fortelle. (2018). “A reinforcement learning based approach for automated lane change maneuvers”. In: *2018 IEEE Intelligent Vehicles Symposium (IV)*. IEEE. 1379–1384.
- Wang, X., Z. Zhu, W. Xu, Y. Zhang, Y. Wei, X. Chi, Y. Ye, D. Du, J. Lu, and X. Wang. (2023a). “OpenOccupancy: A Large Scale Benchmark for Surrounding Semantic Occupancy Perception”. *arXiv preprint arXiv:2303.03991*.
- Wang, Y., Z. Wang, K. Han, P. Tiwari, and D. B. Work. (2021b). “Personalized Adaptive Cruise Control via Gaussian Process Regression”. In: *2021 IEEE International Intelligent Transportation Systems Conference (ITSC)*. IEEE. 1496–1502.
- Wang, Y., C. Huang, Z. Wang, Z. Wang, and Q. Zhu. (2022a). “Design-while-verify: correct-by-construction control learning with verification in the loop”. In: *Proceedings of the 59th ACM/IEEE Design Automation Conference*. 925–930.
- Wang, Y., C. Huang, Z. Wang, S. Xu, Z. Wang, and Q. Zhu. (2021c). “Cocktail: Learn a better neural network controller from multiple experts via adaptive mixing and robust distillation”. In: *2021 58th ACM/IEEE Design Automation Conference (DAC)*. IEEE. 397–402.
- Wang, Y., C. Huang, and Q. Zhu. (2020c). “Energy-efficient control adaptation with safety guarantees for learning-enabled cyber-physical systems”. In: *Proceedings of the 39th International Conference on Computer-Aided Design*. 1–9.

- Wang, Y., S. Zhan, Z. Wang, C. Huang, Z. Wang, Z. Yang, and Q. Zhu. (2023b). “Joint differentiable optimization and verification for certified reinforcement learning”. In: *Proceedings of the ACM/IEEE 14th International Conference on Cyber-Physical Systems (with CPS-IoT Week 2023)*. 132–141.
- Wang, Y., S. S. Zhan, R. Jiao, Z. Wang, W. Jin, Z. Yang, Z. Wang, C. Huang, and Q. Zhu. (2023c). “Enforcing hard constraints with soft barriers: Safe reinforcement learning in unknown stochastic environments”. In: *ICML*. PMLR.
- Wang, Y., W. Zhou, J. Fan, Z. Wang, J. Li, X. Chen, C. Huang, W. Li, and Q. Zhu. (2023d). “POLAR-Express: Efficient and Precise Formal Reachability Analysis of Neural-Network Controlled Systems”. accepted by the *IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems (TCAD)*.
- Wang, Y. (2012). “Gauss–Newton method”. *WIREs Computational Statistics*. 4: 415–420.
- Wang, Z., Y. Bian, S. E. Shladover, G. Wu, S. E. Li, and M. J. Barth. (2020d). “A Survey on Cooperative Longitudinal Motion Control of Multiple Connected and Automated Vehicles”. *IEEE Intelligent Transportation Systems Magazine*. 12(1): 4–24.
- Wang, Z., X. Liao, C. Wang, D. Oswald, G. Wu, K. Boriboonsomsin, M. Barth, K. Han, B. Kim, and P. Tiwari. (2020e). “Driver Behavior Modeling using Game Engine and Real Vehicle: A Learning-Based Approach”. *IEEE Transactions on Intelligent Vehicles*: 1–1.
- Wang, Z., G. Wu, and M. J. Barth. (2020f). “Cooperative Eco-Driving at Signalized Intersections in a Partially Connected and Automated Vehicle Environment”. *IEEE Transactions on Intelligent Transportation Systems*. 21(5): 2029–2038.
- Wang, Z., C. Huang, H. Kim, W. Li, and Q. Zhu. (2021d). “Cross-Layer Adaptation with Safety-Assured Proactive Task Job Skipping”. *ACM Trans. Embed. Comput. Syst.* 20(5s). doi: [10.1145/3477031](https://doi.org/10.1145/3477031).
- Wang, Z., C. Huang, Y. Wang, C. Hobbs, S. Chakraborty, and Q. Zhu. (2021e). “Bounding Perception Neural Network Uncertainty for Safe Control of Autonomous Systems”. In: *DATE’21: Proceedings of the Conference on Design, Automation and Test in Europe*.

- Wang, Z., C. Huang, and Q. Zhu. (2022b). “Efficient Global Robustness Certification of Neural Networks via Interleaving Twin-Network Encoding”. In: *DATE’22: Proceedings of the Conference on Design, Automation and Test in Europe*.
- Wang, Z., H. Liang, C. Huang, and Q. Zhu. (2020g). “Cross-layer design of automotive systems”. *IEEE Design & Test*. 38(5): 8–16.
- Wei, H., G. Zheng, H. Yao, and Z. Li. (2018). “Intellilight: A reinforcement learning approach for intelligent traffic light control”. In: *Proceedings of the 24th ACM SIGKDD International Conference on Knowledge Discovery & Data Mining*. 2496–2505.
- Wei, Y., L. Zhao, W. Zheng, Z. Zhu, J. Zhou, and J. Lu. (2023). “SurroundOcc: Multi-Camera 3D Occupancy Prediction for Autonomous Driving”. *arXiv preprint arXiv:2303.09551*.
- Weidner, W., F. W. Transchel, and R. Weidner. (2017). “Telematic driving profile classification in car insurance pricing”. *Annals of actuarial science*. 11(2): 213–236.
- Wu, T., S. Liu, B. Yu, S. Wang, Y. Bao, and W. Shi. (2022). “INTERNEURON: A Middleware with Multi-Network Communication Reliability for Infrastructure Vehicle Cooperative Autonomous Driving”. *arXiv preprint arXiv:2210.15939*.
- Wu, T., B. Wu, S. Wang, L. Liu, S. Liu, Y. Bao, and W. Shi. (2021). “Oops! It’s Too Late. Your Autonomous Driving System Needs a Faster Middleware”. *IEEE Robotics and Automation Letters*. 6(4): 7301–7308.
- Xie, J., J. Tang, Y. Wang, Q. Zhu, and S. Liu. (2021). “Brief Industry Paper: An Infrastructure-Aided High Definition Map Data Provisioning Service for Autonomous Driving”. In: *2021 IEEE 27th Real-Time and Embedded Technology and Applications Symposium (RTAS)*. IEEE. 421–424.
- Xie, W., T. Hu, N. Ling, G. Xing, S. Liu, and G. Nan. (2023). “TIMELY FUSION OF SURROUND RADAR/LIDAR FOR OBJECT DETECTION IN AUTONOMOUS DRIVING SYSTEMS”. In: *2023 Design, Automation & Test in Europe Conference & Exhibition (DATE)*. IEEE. 1–6.

- Xiong, X., Y. Liu, T. Yuan, Y. Wang, Y. Wang, and H. Zhao. (2023). “Neural Map Prior for Autonomous Driving”. In: *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition*. 17535–17544.
- Xu, Z., X. Li, X. Zhao, M. H. Zhang, and Z. Wang. (2017). “DSRC versus 4G-LTE for Connected Vehicle Applications: A Study on Field Experiments of Vehicular Communication Performance”. *Journal of Advanced Transportation*.
- Yan, C., W. Xu, and J. Liu. (2016). “Can you trust autonomous vehicles: Contactless attacks against sensors of self-driving vehicle”. *DEFCON*.
- Yan, Z., K. Yang, Z. Wang, B. Yang, T. Kaizuka, and K. Nakano. (2019). “Time to lane change and completion prediction based on Gated Recurrent Unit Network”. In: *2019 IEEE Intelligent Vehicles Symposium (IV)*. IEEE. 102–107.
- Yang, B., R. Guo, M. Liang, S. Casas, and R. Urtasun. (2020). “Radar-net: Exploiting radar for robust perception of dynamic objects”. In: *Computer Vision–ECCV 2020: 16th European Conference, Glasgow, UK, August 23–28, 2020, Proceedings, Part XVIII 16*. Springer. 496–512.
- Yang, T. T., T. T. Yang, A. Liu, J. Tang, N. An, S. Liu, and X. Liu. (2022). “AICOM-MP: an AI-based Monkeypox Detector for Resource-Constrained Environments”. *arXiv preprint arXiv:2211.14313*.
- Yao, L., H. Zhao, J. Tang, S. Liu, and J.-L. Gaudiot. (2021). “Streaming Data Priority Scheduling Framework for Autonomous Driving by Edge”. In: *2021 IEEE 45th Annual Computers, Software, and Applications Conference (COMPSAC)*. IEEE. 37–42.
- Yu, B., C. Chen, J. Tang, S. Liu, and J.-L. Gaudiot. (2022). “Autonomous Vehicles Digital Twin: A Practical Paradigm for Autonomous Driving System Development”. *Computer*. 55(9): 26–34.
- Yu, B., W. Hu, L. Xu, J. Tang, S. Liu, and Y. Zhu. (2020). “Building the computing system for autonomous micromobility vehicles: Design constraints and architectural optimizations”. In: *2020 53rd Annual IEEE/ACM International Symposium on Microarchitecture (MICRO)*. IEEE. 1067–1081.

- Yu, B., J. Tang, and S. Liu. (2021). “On designing computing systems for autonomous vehicles: A perceptin case study”. In: *Proceedings of the 26th Asia and South Pacific Design Automation Conference*. 742–747.
- Yu, B., J. Tang, and S. Liu. (2023). “Invited: Autonomous Driving Digital Twin Empowered Design Automation: A Industry Perspective”. In: *2023 60th ACM/IEEE Design Automation Conference (DAC)*. IEEE.
- Yu, H., H. E. Tseng, and R. Langari. (2018). “A human-like game theory-based controller for automatic lane changing”. *Transportation Research Part C: Emerging Technologies*. 88: 140–158.
- Zhang, M., C. Chen, T. Wo, T. Xie, M. Z. A. Bhuiyan, and X. Lin. (2017). “SafeDrive: online driving anomaly detection from large-scale vehicle data”. *IEEE Transactions on Industrial Informatics*. 13(4): 2087–2096.
- Zhang, Q., S. Hu, J. Sun, Q. A. Chen, and Z. M. Mao. (2022). “On adversarial robustness of trajectory prediction for autonomous vehicles”. In: *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition*. 15159–15168.
- Zhang, Z., S. Liu, G. Tsai, H. Hu, C.-C. Chu, and F. Zheng. (2018). “Pirvs: An advanced visual-inertial slam system with flexible sensor fusion and hardware co-design”. In: *2018 IEEE International Conference on Robotics and Automation (ICRA)*. IEEE. 3826–3832.
- Zhao, H., J. Gao, T. Lan, C. Sun, B. Sapp, B. Varadarajan, Y. Shen, Y. Shen, Y. Chai, and C. Schmid. (2021a). “Tnt: Target-driven trajectory prediction”. In: *Conference on Robot Learning*. PMLR. 895–904.
- Zhao, P., W. Niu, G. Yuan, Y. Cai, H.-H. Sung, S. Liu, S. Liu, X. Shen, B. Ren, and Y. Wang. (2021b). “Brief industry paper: Towards real-time 3D object detection for autonomous vehicles with pruning search”. In: *2021 IEEE 27th Real-Time and Embedded Technology and Applications Symposium (RTAS)*. IEEE. 425–428.

- Zhao, X., X. Liao, Z. Wang, G. Wu, M. J. Barth, K. Han, and P. Tiwari. (2021c). “Co-Simulation Platform for Modeling and Evaluating Connected and Automated Vehicles and Human Behavior in Mixed Traffic”. *SAE MobilityRxiv™ Preprint*. DOI: <https://doi.org/10.47953/SAE-PP-00206>.
- Zhao, Y., H. Zhu, R. Liang, Q. Shen, S. Zhang, and K. Chen. (2019). “Seeing isn’t believing: Towards more robust adversarial attack against real world object detectors”. In: *CCS*.
- Zhao, Z., Z. Wang, K. Han, R. Gupta, P. Tiwari, G. Wu, and M. J. Barth. (2022). “Personalized car following for autonomous driving with inverse reinforcement learning”. In: *2022 International Conference on Robotics and Automation (ICRA)*. IEEE. 2891–2897.
- Zheng, B., H. Liang, Q. Zhu, H. Yu, and C. W. Lin. (2016a). “Next Generation Automotive Architecture Modeling and Exploration for Autonomous Driving”. In: *2016 IEEE Computer Society Annual Symposium on VLSI (ISVLSI)*. 53–58. DOI: [10.1109/ISVLSI.2016.126](https://doi.org/10.1109/ISVLSI.2016.126).
- Zheng, B., C.-W. Lin, H. Liang, S. Shiraishi, W. Li, and Q. Zhu. (2017a). “Delay-aware design, analysis and verification of intelligent intersection management”. In: *2017 IEEE International Conference on Smart Computing (SMARTCOMP)*. IEEE. 1–8.
- Zheng, B., C.-W. Lin, S. Shiraishi, and Q. Zhu. (2019). “Design and analysis of delay-tolerant intelligent intersection management”. *ACM Transactions on Cyber-Physical Systems*. 4(1): 1–27.
- Zheng, B., C.-W. Lin, H. Yu, H. Liang, and Q. Zhu. (2016b). “CONVINCE: A cross-layer modeling, exploration and validation framework for next-generation connected vehicles”. In: *2016 IEEE/ACM International Conference on Computer-Aided Design (ICCAD)*. ACM. 1–8.
- Zheng, S., Y. Yue, and P. Lucey. (2017b). “Generating long-term trajectories using deep hierarchical networks”. *arXiv preprint arXiv:1706.07138*.
- Zhu, Q., C. Huang, R. Jiao, S. Lan, H. Liang, X. Liu, Y. Wang, Z. Wang, and S. Xu. (2021). “Safety-assured design and adaptation of learning-enabled autonomous systems”. In: *Proceedings of the 26th Asia and South Pacific Design Automation Conference*. 753–760.

- Zhu, Q., W. Li, H. Kim, Y. Xiang, K. Wardega, Z. Wang, Y. Wang, H. Liang, C. Huang, J. Fan, and H. Choi. (2020). “Know the Unknowns: Addressing Disturbances and Uncertainties in Autonomous Systems”. In: *Proceedings of the 39th International Conference on Computer-Aided Design. ICCAD '20*. Virtual Event, USA. doi: [10.1145 / 3400302.3415768](https://doi.org/10.1145/3400302.3415768).
- Zhu, Q. and A. Sangiovanni-Vincentelli. (2018). “Codesign Methodologies and Tools for Cyber–Physical Systems”. *Proceedings of the IEEE*. 106(9): 1484–1500. doi: [10.1109/JPROC.2018.2864271](https://doi.org/10.1109/JPROC.2018.2864271).