

The Journey from Powertrain Control to Emerging Mobility Systems

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My research interests span several fields, including analysis, optimization, and control of cyber-physical (CPS) systems; decentralized stochastic systems, stochastic scheduling, and resource allocation. The overarching goal of my research is to establish a rigorous theoretical framework aimed at enhancing our understanding of the behavior of large scale, complex CPS systems and develop decentralized control algorithms for making such systems able to learn to improve their performance over time while interacting with their environment. As we move to increasingly complex cyber-physical systems with an expanded feature space, fundamentally new approaches are needed to understand the impact on system behavior of the interplay between subsystems of different physical processes, at different scales, or between decision points in an engineered system. The emphasis is on applications related to connected and automated vehicles (CAVs), smart cities, and sociotechnical systems. My interest in developing control algorithms that could make systems able to learn their optimal operation started early on, while I was still at graduate school, when I read an article about the discrepancy between true fuel economy of a vehicle and the one posted on the window sticker. The article was discussing the implications of the driver's driving style on engine operation, and stated that the state-of-the-art control methods, by that time, consist of static controllers which cannot optimize engine operation for different driving styles but only for predetermined ones. This article provided inspiration that eventually led to forming the topic of my dissertation. In my dissertation [1], I developed the theoretical framework [2–5] and control algorithms [6–8] that can turn the engine of a vehicle into an autonomous intelligent system capable of learning its optimal operation in real time while the driver is driving the vehicle. I modeled the evolution of the state of the engine as a control Markov chain [9] and proved [10] that it eventually converges to a stationary probability distribution deemed characteristic of the driver's driving style. Through this approach, the engine progressively perceives the driver's driving style [11] and eventually learns to operate in a manner that optimizes specified performance criteria, e.g., fuel economy, emissions with respect to the driver's driving style. The framework also allows the engine to identify the driver, and thus it can adjust its operation to be optimal for any driver based on what it has learned in the past regarding her/his driving style. The outcome of my dissertation research eventually led to a US patent [12].

Moving to General Motors Research & Development as a Senior Researcher, I had the chance to continue working on self-learning control for advanced powertrain systems. I led several projects on autonomous intelligent propulsion systems and developed computational mathematical models and control algorithms towards making highly energy-efficient and eco-friendly vehicles. I was a member of the team that demonstrated successfully the implementation of self-learning control algorithms [13] in two demo vehicles, Saturn Aura and Opel Vectra.

When I joined Oak Ridge National Laboratory (ORNL) as an Alvin M. Weinberg Fellow, although the focus of my fundamental research interests remained the same, the emphasis of the applications shifted from powertrain systems to vehicles, and then to CAVs. At ORNL, I had the chance to work across different technical areas including stochastic optimal control [14–16], optimal design and power management control and routing of hybrid electric vehicles (HEVs) and plug-in HEVs (PHEVs) [17–27], and driver's feedback systems [28–30]. The latter eventually led to a technology [31] that was licensed in SanTed Project Management LLC. I also contributed to the solution of problems that included smart buildings aimed at optimizing energy system parameters to (1) improve sustainability, (2) facilitate cost-effective energy generation, and (3) allocate demand optimally to different energy sources, e.g., solar, renewable, etc [32–34]. On the fundamental research front, I established the theoretical framework for the analysis and stochastic control

of complex systems consisting of interactive subsystems [35]. In particular, I developed a duality framework and showed that the Pareto control policy minimizes the long-run expected average cost criterion of the system while also presented a geometric interpretation of the solution and conditions for its existence. I provided theoretical results showing that the Pareto control policy provides an equilibrium operating point among the subsystems, and if the system operates at this equilibrium, then the long-run expected average cost per unit time is minimized. This result implies that the Pareto control policy can be of value when we seek to derive the optimal control policy for complex systems online. Later on, and in my role as the Deputy Director of the Urban Dynamics Institute at ORNL, I developed several initiatives with the goal to investigate how we can use scalable data and informatics to enhance understanding of the environmental implications of CAVs and improve transportation sustainability and accessibility. I contributed towards the development of a decentralized optimal control framework whose closed-form solution exists under certain conditions, and which, based on Hamiltonian analysis, yields for each vehicle the optimal acceleration/deceleration, in terms of fuel consumption. The solution allows the vehicles to cross merging roadways without creating congestion, and under the hard safety constraint of collision avoidance [36–40].

When I joined the University of Delaware, I established the Information and Decision Science (IDS) Laboratory with the vision to advance the state of the art in the analysis, optimization, and control of cyber-physical (CPS) networks. The overarching goal of the IDS Lab is to enhance our understanding of the behavior of large scale, complex CPS networks consisting of multiple entities (or agents). The emphasis is on applications related to emerging mobility systems and sociotechnical systems. Emerging mobility systems are typical CPS networks where the cyber component (e.g., data and shared information through vehicle-to-vehicle and vehicle-to-infrastructure communication) can aim at optimally controlling the physical entities (e.g., CAVs, non-CAVs). A mobility system encompasses the interactions of three heterogeneous dimensions: (1) transportation systems and modes, e.g., CAVs, shared mobility, and public transit integrated with advanced control algorithms, (2) social behavior of drivers, operators (for autonomous vehicles), and travelers (or pedestrians) interacting with these systems, and (3) information management of data available and shared information. The constellation of these three dimensions constitutes a sociotechnical system that should be analyzed holistically. The CPS nature of emerging mobility systems is associated not only with technological and information management dimensions but also with human adoption (social dimension). My students and I, in conjunction with my collaborators, have made contributions on the technological dimension of mobility systems by developing control algorithms for optimal coordination of CAVs [41–72] and identifying potential research paths with connected autonomous systems [73]. However, I came to realize that current methods analyze, design, and optimize a mobility system without considering the social dimension resulting in systems that might not be acceptable by the drivers, travelers, and the public. In particular, one key research question that still remains unanswered is “how can we develop an energy-efficient mobility system that can be widely acceptable by drivers, travelers, and the public?” To address this question, my students and I are taking the following research steps that combine the three aforementioned dimensions [74–76]: (1) explore how advanced control technologies in conjunction with Big Data from vehicles and infrastructure can improve the efficiency of transportation systems and modes, e.g., eliminate stop-and-go driving, reduce congestion; (2) investigate public attitudes toward emerging transportation systems and identify the human behavioral and emotional responses to systems such as CAVs and shared mobility, and (3) address the negative rebound effects of improving the efficiency of transportation systems by exploring whether household activities and travel demand might increase if the efficiency of the transportation systems improves. Step 1 will identify the new congestion patterns of optimized transportation systems and modes. Step 2 will examine public reaction, adoption, and use of a potential energy-efficient mobility system, which will determine the urban planning, public policy,

and governance frameworks to enable the system-wide optimal outcomes. Step 3 will determine the new levels of travel demand and, eventually, the impact on vehicle miles traveled. The expected outcome of my group's research in this area will aim at identifying a mobility system which is not only energy efficient but also acceptable by the drivers, travelers, and the public.

Another research direction that my students and I are currently working is towards establishing a theoretical framework and control algorithms to enhance our understanding of flocking [77–82] and the behavior of large scale, decentralized CPS networks [83–86]. One common characteristic in CPS networks is that the models used cannot predict their behavior. In particular, one research question that still remains unanswered is “how can we establish a framework aimed at both predicting and affecting the behavior of a CPS network?” To address this question, my students and I are taking the following research steps: (1) explore the communication structure of CPS networks; (2) investigate how data in a CPS network that is increasing with time can be “compressed” to a sufficient statistic taking values in a time-invariant space (structural results); (3) implement a desired emerging behavior in a class of CPS networks involving multiple self-interested agents, each with private information and preferences. Step 1 will provide the modeling framework of the information structure of CPS networks which is necessary to understand how information is propagated within different classes of CPS networks. Step 2 will provide the structural results required for implementing a decentralized control framework to control CPS networks. Step 3 will aim at imposing the conditions among the agents under which a CPS network will exhibit the desired behavior. The expected outcome and foundation of my group's research in this area will aim at enhancing our understanding of the behavior of large scale, complex CPS networks and develop, networking capabilities to analyze, model, simulate, and predict complex phenomena.

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