SCHEDULING

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Abstract. In this article, we discuss the optimization of a linearized traffic flow network model based on conservation laws. We present two solution approaches. One relies on the classical Lagrangian formalism (or adjoint calculus), whereas another one uses a discrete mixed-integer framework. We show how both approaches are related to each other theoretically. Numerical experiments are accompanied to show the quality of solutions.

Key words. traffic networks, network optimization

AMS subject classifications. 90B20, 90C11

1. Extended Abstract.

1.1. problem and objectives.

- assemble to order for capacity planning
- 3%, 5%, 3 min C&R-App

1.2. literature.

- TAKT
- OpSysTra

1.3. innovation.

- first infrastructure manager to use optimization models for freight train timetable planning
- customer value: shorter travel time, higher infrastructure capacity and faster response time
- first fully automated process in short-term capacity planning from ordering to the actual departure of the train

1.4. data.

- digital infrastructure model
- timetable of passenger trains

1.5. practical relvance.

- release of C&R-App in 2019
- first step in automization of timetable planning and traffic management
- prepares DB Netz for the future with an increased efficiency (higher capacity and less effort for staff and customers)
- **2.** Introduction. Modeling, simulation and optimization of traffic flow networks based on partial differential equations have been investigated intensively during the last years, see for instance for an overview [4, 11, 15, 16, 17, 19, 21].

For optimization purposes, different applications such as optimal routing of traffic at intersections [11, 15, 16], traffic light control [12], evacuation planning [13, 14] and air traffic control [1] are of interest. Since in all problems the underlying optimization problem is restricted by nonlinear partial differential equations, relaxed models with simplified dynamics have been investigated. In this context, two different solution

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approaches emerge. On the one hand, nonlinear continuous optimization techniques have been successfully applied. To compute the optimum, the first order optimality system is derived and solved by a descent type method, see [18, 22, 25]. On the other hand, suitable discretization of the dynamics leads to linear network flow models that have been widely considered in the field of combinatorial optimization [2, 8, 9, 20].

In fact, there exist a few research results comparing both optimization tools, see [5, 7, 10, 11, 26]. One might assume that a one-to-one relation between nonlinear continuous and discrete optimization techniques holds true. This is especially the case when the governing dynamics in the network are linear. Then, appropriate numerical discretizations can be chosen, such that the original optimization problem either leads to numerically solving the finite-dimensional optimality system, i.e. the so-called discretize-then-optimize approach, or the interpretation as a mixed-integer programming model (MIP). However, it remains the question of detecting local or global optima. We know from the theory of linear programming, in particular the strong duality theorem [3, 23, 24], that under certain circumstances a global optimum can be reached. This is usually not the case for the adjoint calculus. The solution of the optimality system via gradient methods often gets stuck in local optima.

In this work we want to close the gap between the two solution procedures that first appear to be quite varied. To do so, we start with the traffic flow network model where the evolution of traffic density on roads is governed by the linearized Lighthill-Whitham-Richards (LWR) equations, see [6]. For the coupling conditions at the intersections, we stick to the ones presented by Coclite-Garavello-Piccoli [4] but do not use a predefined distribution matrix since these parameters will be obtained as solutions of the optimization problem. In a next step, we discretize the full network model in space and time and formally derive the adjoint equations and the mixed-integer model (MIP) as well. We show the equivalence of both approaches by comparing the dual variables of the MIP and the adjoint variables. A formal, detailed proof for a simple network is given. Finally, we compare the optimization results of the MIP and the adjoint equations for different networks and objective functions numerically.

The paper is organized as follows: Section 3 contains a description of the network and the models used for describing flow on roads. Section 4 introduces the optimization problem to be considered. Herein, we also present the main theorem of this article, which we prove in the following sections. In section ??, the adjoint equations for the former optimization problem are derived. Thereafter, the mixed-integer model is introduced in section 5. In section ??, we conclude the proof of our main theorem. Numerical examples are given in section 6.

- 3. Traffic Flow Network Model.
- 4. Optimization Problem.
- 5. Mixed Integer Program (MIP).
- 6. Numerics.

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SCHEDULING 3

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