

On the influence of time delays in vehicular traffic

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Outline

- 1 Framework and motivations
- 2 Delayed LWR model and its theoretical properties
 - Formal derivation
 - Conservation - Positivity - Boundedness
- 3 Numerical discretization and its properties
- 4 Numerical tests
 - Key example: Stop & Go waves
- 5 Second order traffic models with time delay

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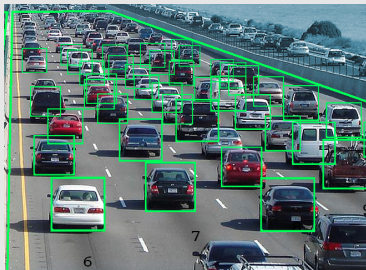
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Framework

Framework

Mathematical modeling of traffic flow on a **single road**, by means of both

- **microscopic** (agent-based) follow-the-leader models based on a system of ODEs
- **MACROSCOPIC** (fluid-dynamic) models based on conservation laws



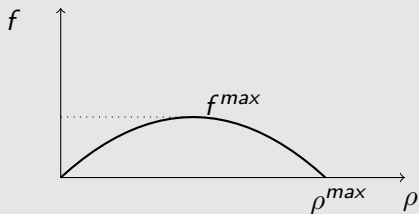
Framework: Macroscopic Model

$\rho(x, t)$ density of cars at point x and time t

$V(x, t)$ velocity of cars at point x and time t

$f(x, t) = \rho(x, t)V(x, t)$ flux of cars at point x and time t

The **fundamental diagram** establishes the relationship between the flux and the density of vehicles, i.e. $\{(\rho(x, t), f(x, t)) : x \in \mathbb{R}, t > 0\}$



Classical LWR Model

Lighthill-Whitham-Richards (LWR) model (1955)

$$\partial_t \rho(x, t) + \partial_x (\rho(x, t) V(\rho(x, t))) = 0, \quad x \in \mathbb{R}, \quad t > 0$$

Main features

- is a hyperbolic conservation law (the total mass has to be preserved)
- the velocity depends on the density and typically
$$V(\rho) = V_{\max} \left(1 - \frac{\rho}{\rho_{\max}} \right)$$
- easy to implement and very cheap in terms of memory and time
- can be used also on large networks for traffic forecast
- **BUT** accelerations are considered to be instantaneous and traffic is described **only** at the equilibrium (no typical features of traffic dynamics as capacity drop or spontaneous congestions like Stop & Go waves)

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Derivation of the delayed LWR model

Question

Is there a significant impact of time delays in (macroscopic) traffic models?

- Consider a delayed microscopic model by Newell¹

$$\dot{x}_i(t) = W\left(\frac{\Delta x_i(t - T)}{\Delta X}\right), \quad i = 1, \dots, N \quad (1)$$

where $W(\cdot)$ is a velocity function, $T > 0$ a reaction time, $\Delta x_i(t) = x_{i+1} - x_i$ the spacing between vehicle i and $i + 1$ and $\Delta X > 0$ a space scaling (i.e. average length of a vehicle)

- Goal:** derive a LWR type model with explicit delay dependence

¹Newell, Oper. Res., 1961

Derivation of the delayed LWR model

- Define the traffic density ρ as the inter-vehicle-spacing²

$$\rho_i(t) = \frac{\Delta X}{\Delta x_i(t - T)}$$

- Rewriting in terms of V and inserting into (1)

$$\dot{x}_i(t) = W\left(\frac{1}{\rho_i(t - T)}\right) = V(\rho_i(t - T))$$

- To link the microscopic and macroscopic description, we consider

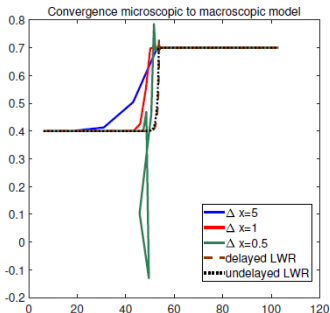
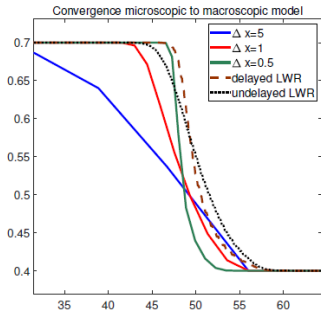
$$\partial_t \frac{1}{\rho_i(t)} = \partial_t \frac{\Delta x_i(t)}{\Delta X} = \frac{V(\rho_{i+1}(t - T)) - V(\rho_i(t - T))}{\Delta X}$$

- Limit process $N \rightarrow \infty$ leads to $\partial_t \frac{1}{\rho(y, t)} = \partial_y V(\rho(y, t - T))$ or

$$\partial_t \rho(x, t) + \partial_x (\rho(x, t) V(\rho(x, t - T))) = 0$$

²Aw, Klar, Materne, Rascle, SIAP, 2002

Microscopic vs. macroscopic approach

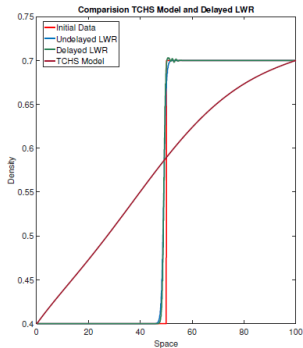
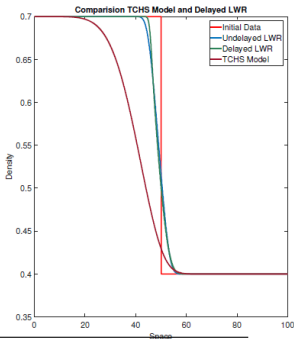


- Two special scenarios: rarefaction wave (left) and shock wave (right)
- Convergence of the microscopic model for ΔX small enough can be observed
- Since the microscopic model is NOT collision free, the model breaks down for $\Delta X = 0.5$

Comments on the delayed LWR model

- A convection-diffusion equation³ can be also derived from (1) by using a Taylor expansion in the argument of W :

$$\partial_t \rho + \partial_x (\rho V(\rho)) = -T \partial_x ((\rho V'(\rho))^2 \partial_x \rho) \quad (\text{TCHS})$$



³Tordeux, Costeseque, Herty, Seyfried, SIAP, 2018

Delayed LWR model

Delayed LWR model

$$\begin{cases} \partial_t \rho(x, t) + \partial_x (\rho(x, t) V(\rho(x, t - T))) = 0 & x \in \mathbb{R}, t \in [0, T_f] \\ \rho(x, t) = \rho^0(x) & x \in \mathbb{R}, t \in [-T, 0] \end{cases}$$

- $T > 0$ is the time delay, which represents the reaction time of both drivers and vehicles
- **Note:** In order to guarantee the well-posedness of the problem, an initial history function as initial data has to be defined on $[-T, 0]$
- In the limit case of $T = 0$ the classical LWR model is recovered
- If the delay is not chosen appropriately, the maximum principle can be violated (ρ can be greater than 1)

Theoretical properties

Lemma - Conservation of mass

The delayed LWR model preserves the quantity $\rho(x, t)$.

⁴Keimer, Pflug, NoDEA, 2019

Theoretical properties

Lemma - Conservation of mass

The delayed LWR model preserves the quantity $\rho(x, t)$.

Lemma - Positivity

Assume we have initial data with non-negative density $\rho(x, t) \geq 0$. Then, for the delayed LWR model the density stays non-negative.

⁴Keimer, Pflug, NoDEA, 2019

Theoretical properties

Lemma - Conservation of mass

The delayed LWR model preserves the quantity $\rho(x, t)$.

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Lemma - Unboundedness

Assume V is monotone decreasing and $V(\rho_{\max}) = 0$ for ρ_{\max} , the maximal density in the classical LWR model. The delayed first order model has no maximal density ρ_{\max} .

Note: In general $\rho > \rho_{\max}$ is allowed. Moreover, existence and uniqueness of the solution needs to be considered carefully⁴

⁴Keimer, Pflug, NoDEA, 2019

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Numerical discretization

Idea: Investigate properties from a numerical viewpoint

- $\Delta x, \Delta t > 0$ define a grid in space $\{x_i = i\Delta x, i \in \mathbb{Z}\}$ and time $\{t^n = n\Delta t, n \in \mathbb{N}\}$.
- $\rho_i^n = \rho(x_i, t^n)$ is the discretized variable
- classical CFL condition: $\Delta t \leq \frac{\Delta x}{\max_k(\lambda_k)}$, where λ_k are the eigenvalues of the Jacobian matrix of f

Lax-Friedrichs Method

$$\rho_i^{n+1} = \frac{1}{2}(\rho_{i+1}^n + \rho_{i-1}^n) - \frac{\Delta t}{2\Delta x}(f(\rho_{i+1}^n) - f(\rho_{i-1}^n))$$

BUT we are dealing with the delayed model...

Numerical scheme

Altered Lax-Friedrichs method

$$\rho_i^{n+1} = \frac{1}{2}(\rho_{i+1}^n + \rho_{i-1}^n) - \frac{\Delta t}{2\Delta x}(f(\rho_{i+1}^{n-T_\Delta}, \rho_{i+1}^n) - f(\rho_{i-1}^{n-T_\Delta}, \rho_{i-1}^n))$$

- $T \geq \Delta t$ to be able to treat with the delay
- T_Δ is the number of steps that make up the time delay T
- Identify flux as $f(\rho_{i+1}^{n-T_\Delta}, \rho_{i+1}^n) = \rho_{i+1}^n V(\rho_{i+1}^{n-T_\Delta})$
- For the well-posedness of the discrete problem, we have to provide an initial history function as initial data defined on $[-T, 0]$, i.e., $\rho^0(x, 0)$ is assumed to be constant in t for $t \in [-T, 0]$

Properties of the discretization

Conservation

At the discrete level, we have:

$$\Delta x \sum_i \rho_i^{n+1} = \Delta x \sum_i \left(\frac{1}{2}(\rho_{i+1}^n + \rho_{i-1}^n) - \frac{\Delta t}{2\Delta x} (V(\rho_{i+1}^{n-T\Delta})\rho_{i+1}^n - V(\rho_{i-1}^{n-T\Delta})\rho_{i-1}^n) \right)$$

This gives us

$$\Delta x \sum_i \rho_i^{n+1} = \Delta x \sum_i \rho_i^n$$

and therefore conservation of mass.

Properties of the discretization

Positivity

To ensure positivity, we need to guarantee

$$\frac{\Delta t}{2\Delta x} (V(\rho_{i+1}^{n-T_\Delta})\rho_{i+1}^n - V(\rho_{i-1}^{n-T_\Delta})\rho_{i-1}^n) \leq \frac{1}{2}(\rho_{i+1}^n + \rho_{i-1}^n).$$

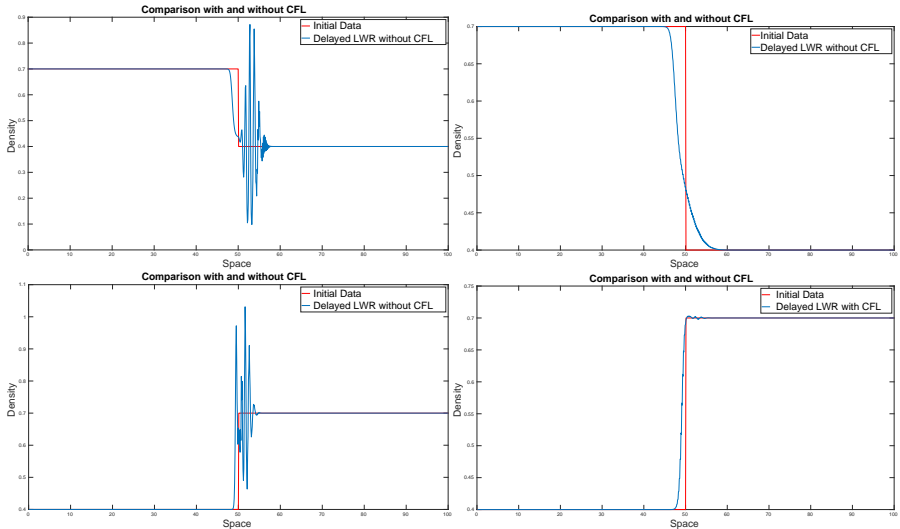
We achieve this result by

- introducing an altered **CFL** condition:

$$\Delta t \leq \frac{\Delta x}{\max\{|\rho^n|, |\rho^{n-T_\Delta}|\}}$$

- assuming $|V(\rho)| \leq \max\{|\rho^n|, |\rho^{n-T_\Delta}|\}$ with $V_{\max} = 1$

Rarefaction (1st row) vs. shock wave (2nd row)



Properties of the discretization

L^∞ bound

Under the altered CFL condition, it holds

$$\|\rho^{n+1}\|_{L^\infty} \leq 2 \max\{|\rho^n|, |\rho^{n-T_\Delta}|\}$$

TV bound

Under the altered CFL condition, it holds

- in space, assuming $TV(\rho_\Delta^n) = \sum_j |\rho_{j+1}^n - \rho_j^n|$:

$$\sum_j |\rho_{j+1}^{n+1} - \rho_j^{n+1}| \leq 2 \left(5 + \frac{1}{\max\{|\rho_j^{n-T_\Delta}|, |\rho_j^n|\}} \right) \max\{TV(\rho_\Delta^n), TV(\rho_\Delta^{n-T_\Delta})\}$$

- in time:

$$\sum_j |\rho_j^{n+1} - \rho_j^n| \leq \sum_j 4 \max\{|\rho_j^{n-T_\Delta}|, |\rho_j^n|\} + 2.$$

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Numerical tests

Starting from empirical observations⁵, we assume the **velocity function** as

$$V(\rho) = \begin{cases} V_{max} & \rho \leq \rho_f \\ \alpha \left(\frac{1}{\rho} - \frac{1}{\rho_c} \right) & \rho_f < \rho < \rho_c \\ 0 & \rho \geq \rho_c \end{cases}$$

- $\alpha > 0$ (parameter)
- $\rho_c \in (0, \rho_{max}]$ represents the so-called *safe distance* at the macroscopic level
- $\rho_f \in [0, \rho_{max})$ represents the distance until vehicles influence each other

Note that $V(\rho)$ **respects** the hypothesis $|V(\rho)| \leq |\rho_{max}|$ as $V_{max} = 1$

⁵ Cristiani, Iacomini, DCDS-B, 2019

Discretization parameters

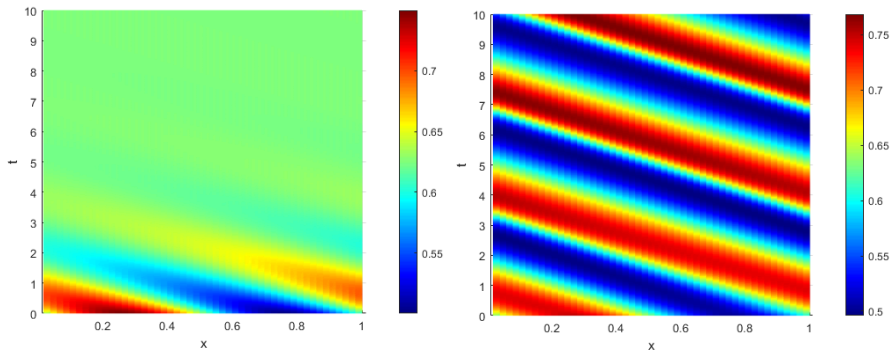
Let us fix the discretization parameters as follows:

- space interval $[a, b] = [0, 1]$, $\Delta x = 0.02$
- periodic boundary conditions
- Δt is chosen such that the CFL condition is satisfied
- $\rho_c = 0.75$ and $\rho_f = 0.2$ as real data suggests⁶
- for simplicity the delay is a multiple of Δt : $T_\Delta \approx (10\Delta t, 20\Delta t)$.

⁶Balzotti, Iacomini, SEMA SIMAI Springer Series, 2020

LWR vs. delayed LWR

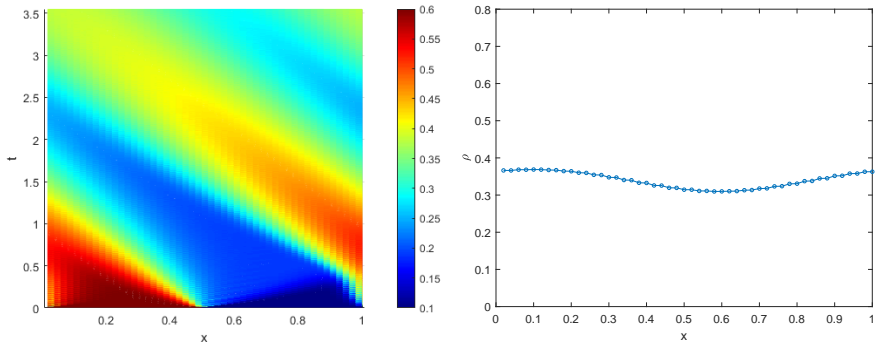
$$\rho^0(x) = \frac{5}{8} + \frac{1}{8} \sin(2\pi x) \quad T_{\Delta} = 15\Delta t$$



➤ Density evolution shows that the **LWR model** smears out the perturbations while the **delayed LWR model** preserves them

The crucial role of the delay: too small

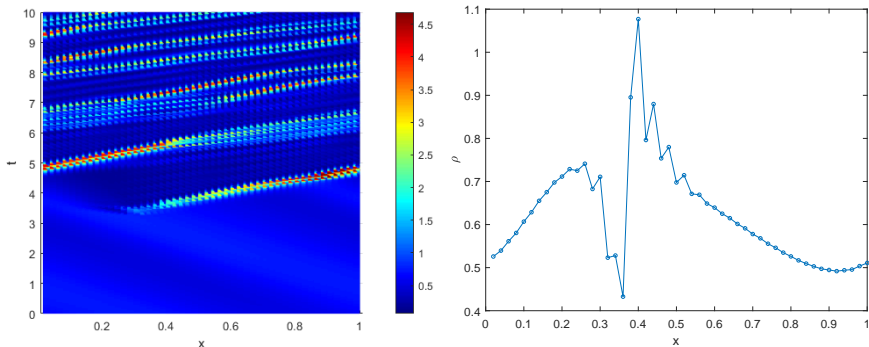
$$\rho^0(x) = \frac{5}{8} + \frac{1}{8} \sin(2\pi x) \quad T_\Delta = 4\Delta t$$



➤ If the delay is too small, we recover a situation similar to the **LWR model**

The crucial role of the delay: too large

$$\rho^0(x) = \frac{5}{8} + \frac{1}{8} \sin(2\pi x) \quad T = 18\Delta t$$



→ The hypothesis are **no longer** satisfied and the density starts to oscillate

Stop & Go waves

Definition

Stop-and-Go waves are a typical feature of congested traffic. A S&G wave is detected when vehicles stop and restart without any apparent reason generating a wave that travels backward with respect to the cars' trajectories.

Relevance

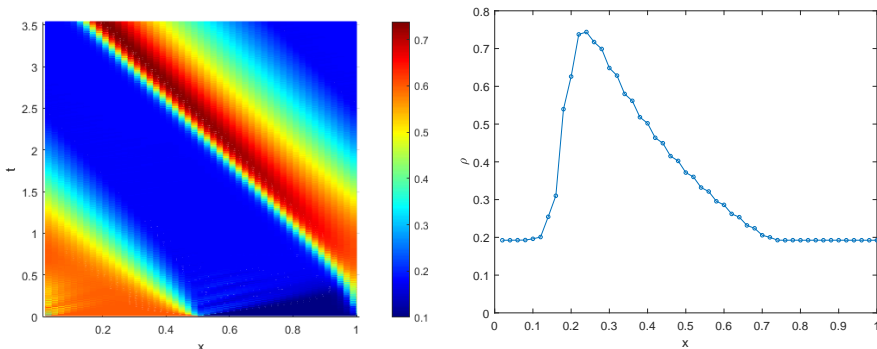
S&G waves are one of the main reasons for

- accidents,
- longer travel times,
- high fuel consumption,
- pollution

→ VIDEO by Stern et al.: <https://youtu.be/2mBjYZTeaTc>

Backward propagation of Stop & Go waves

$$\rho^0(x) = \begin{cases} 0.6 & x < 0.5 \\ 0.1 & x \geq 0.5. \end{cases} \quad T_{\Delta} = 10\Delta t$$

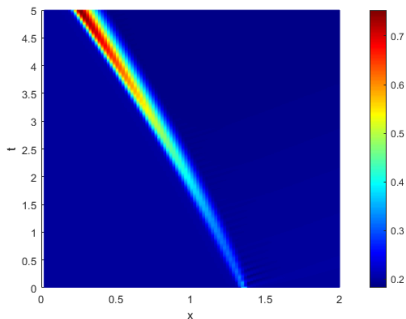


→ Density profile at $T_{\text{final}} = 3.5$ yields a well-defined S&G wave⁷

⁷ Flynn et al., Phys. Rev. E, 2009

Triggering of Stop & Go waves

$$\rho^0(x) = \begin{cases} 0.35 & 1.34 \leq x \leq 1.342 \\ 0.2 & \text{elsewhere,} \end{cases} \quad T = 21\Delta t$$



→ Initial perturbation increases and moves backward as time increases

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A note on delayed second order models

Follow-up question

What do we observe in second order delayed models?

Starting from the model by Gazis, Herman, Rothery (1961)

$$\dot{x}_i(t) = v_i(t)$$

$$\dot{v}_i(t) = C \frac{(v_{i+1}(t-T) - v_i(t-T))}{(x_{i+1}(t-T) - x_i(t-T))^{\gamma+1}}, \quad i = 1, \dots, N,$$

with delay $T > 0$ and model constants $C = v_{\text{ref}} \Delta X^\gamma > 0$ and $\gamma \geq 0$, we derive the **delayed Aw-Rascle-Zhang (ARZ) model**

$$\partial_t \rho(x, t) + \partial_x (\rho(x, t) v(x, t)) = 0 \quad (\text{RSD})$$

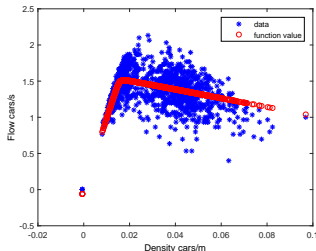
$$\partial_t (\rho(x, t) w(x, t)) + \partial_x (\rho(x, t) v(x, t) w(x, t)) = Q(\partial_x v, \rho)$$

$$\rightsquigarrow Q(\partial_x v, \rho) = v_{\text{ref}} (\partial_x v(x, t-T) \rho(x, t-T)^\gamma - \partial_x v(x, t) \rho(x, t)^\gamma),$$

where $w = v + P(\rho)$ and $P(\rho)$ is a known pressure function

Data-fitting

- Take real data from the Minnesota Department of Transportation⁸



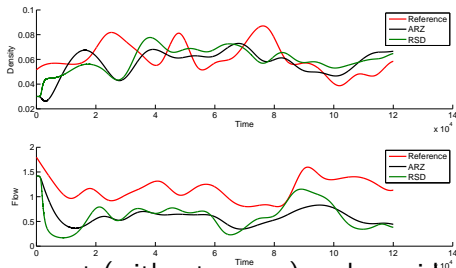
- $Q(\rho)$ fits the fundamental diagram data in a least squares sense⁹:

$$Q(\rho) = \alpha \left[\sqrt{1 + (\lambda p)^2} + \left(\sqrt{1 + (\lambda(1 - p))^2} - \sqrt{1 + (\lambda p)^2} \right) \frac{\rho}{\rho_{\max}} - \sqrt{1 + \lambda^2 \left(\frac{\rho}{\rho_{\max}} - p \right)^2} \right]$$

⁸ <http://data.dot.state.mn.us/datatools/>

⁹ Fan, Herty, Seibold, NHM, 2014

Data-fitting: Comparison of the ARZ to the delayed ARZ model



- Take a road segment (without ramps) and consider the time interval 16:00-16:10 (workday)
- The classical ARZ model tends to avoid deviations
- Error between real data and simulated models:

$$E(x, t) = \frac{|\rho_{\text{model}}(x, t) - \rho_{\text{data}}(x, t)|}{\Delta \rho} + \frac{|v_{\text{model}}(x, t) - v_{\text{data}}(x, t)|}{\Delta v},$$

where $E_{\text{ARZ}} = 0.6404$ and $E_{\text{RSD}} = 0.6034$

Conclusion and future perspective

Conclusions

- introduced new delayed macroscopic traffic models, pointing out similarities and differences to the undelayed versions
- proposed an altered Lax-Friedrichs method to compute the numerical solution
- showed that the LWR delayed model is able to reproduce Stop & Go waves while the delayed ARZ model performs well regarding real data

Future Perspective

- advanced parameter estimation techniques
- inclusion of stochastic events for route planning

This talk is based on the following references:

- M. Burger, S. Göttlich and T. Jung, *Derivation of a first order traffic flow model of Lighthill-Whitham-Richards type*, IFAC PapersOnLine, 51, pp. 49–54, 2018.
- M. Burger, S. Göttlich and T. Jung, *Derivation of second order traffic flow models with time delays*, Netw. Heterog. Media, Vol. 14(2), pp. 265-288, 2019.
- S. Göttlich, E. Iacomini and T. Jung, *Properties of the LWR model with time delay*, to appear in Netw. Heterog. Media, 2020.
- T. Jung, *Delayed Traffic Models in Multiple Scales: New Macroscopic Models and Their Numerics*, PhD Thesis, University of Mannheim, to appear 2020.
- ... and was supported by the German Research Foundation (DFG).

Thank you for your attention!

