

# Lagrangian Resolution of the Multilane Hybrid Traffic Flow Model

[Microscopic modeling of the relaxation phenomenon using a macroscopic lane-changing model. Trans. Res. B, forthcoming]

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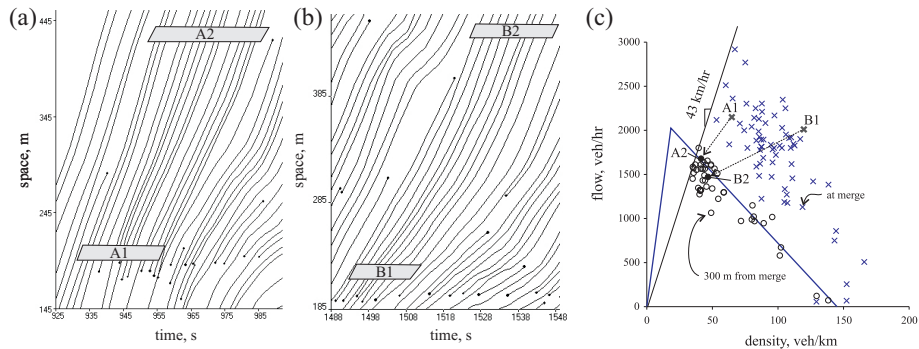
Mathematical Models of Traffic Flow

# Motivation

## Current Lane-Changing models

- tend to be inconsistent
- have too many parameters
  - curtesy
  - forced LCs
- do not address relaxation after the LC

# Relaxation phenomenon



# Outline

- 1 Background
  - The multilane hybrid (MH) model
  - The car-following linear (CFL) model
- 2 Model formulation
  - Quantizing  $\Phi$
  - The microscopic moving bottleneck model
  - The relaxation model
- 3 Model validation

# Outline

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# The multilane hybrid (MH) model

[Laval and Daganzo. Lane-changing in traffic streams. *Trans. Res. B*, 40 (3): 251-264, 2006]

## Definition

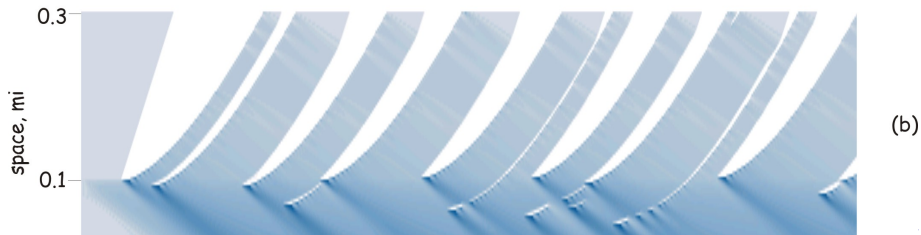
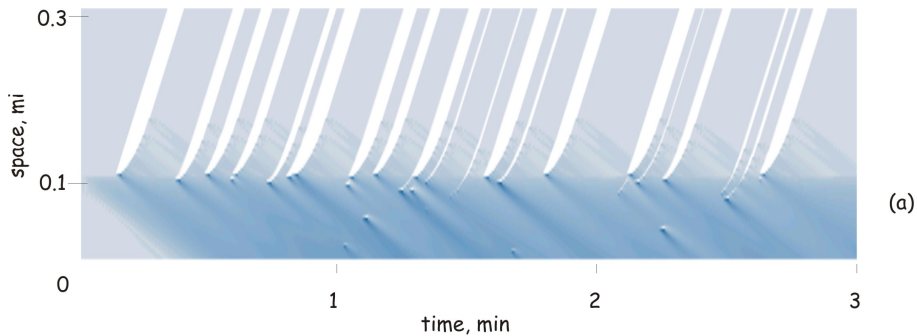
- Conservation law in each lane

$$\frac{\partial k_\ell}{\partial t} + \frac{\partial q_\ell}{\partial x} = \Phi_\ell, \quad \ell = 1 \dots n \quad (1)$$

Where  $\Phi_\ell$  is the *net lane-changing rate onto* lane  $\ell$  in units of veh/time-distance.

- disruptive lane-changes  $\sim \text{Poisson}(\Phi)$
- disruptive lane-changes act as moving bottleneck in target lane

# Moving Bns = key to capacity drop



# The multilane hybrid (MH) model I

- the sending (demand) function

$$\lambda_\ell = \min\{uk_\ell, Q\} \quad (2)$$

- the receiving (supply) function

$$\mu_\ell = \min\{w(\kappa - k_\ell), Q\} \quad (3)$$

- for triangular FD with

- free-flow speed  $u$
- wave speed  $w$ , and
- jam density/lane  $\kappa$

$\Rightarrow Q = uw\kappa/(u + w)$  is the one-lane capacity.

# The multilane hybrid (MH) model II

- The desired number of lane-changing moves

$$L_{\ell\ell'} \Delta x = \lambda_{\ell} \pi_{\ell\ell'} \Delta t \quad \forall \ell, \forall \ell' \neq \ell \quad (4)$$

where

$$\pi_{\ell\ell'} = \frac{\Delta v_{\ell\ell'}}{u\tau} \quad (\text{discretionary LCs}) \quad (5)$$

- the parameter  $\tau$  has units of time.
- the desired number of through moves is

$$T_{\ell'} = \left(1 - \sum_{\ell \neq \ell'} \pi_{\ell'e} \Delta t\right) \lambda_{\ell'} \quad \forall \ell'. \quad (6)$$

## The multilane hybrid (MH) model III

- available capacity is prorated according to demands  $\Rightarrow$  fraction of the demand able to advance to  $\ell'$  is

$$\gamma_{\ell'} = \min\left\{1, \frac{\mu_{\ell'}}{T_{\ell'} + \sum_{\ell \neq \ell'} \Delta x L_{\ell\ell'}}\right\} \quad (7)$$

- and the transfers to the target lane  $\ell$  are

$$\Phi_{\ell\ell'} = \gamma_{\ell'} L_{\ell\ell'}, \quad \forall \ell \neq \ell' \quad (8)$$

$$q_{\ell'} = \gamma_{\ell'} T_{\ell'} \quad (9)$$

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# The car-following linear (CFL) model

- exact car-following formulation of kinematic wave theory with triangular FD for a single lane

$$x_{j+1,\ell}^i = \min\{x_{j\ell}^i + u\Delta t, x_{j\ell}^{i-1} - 1/\kappa\} \quad (10)$$

where  $x_j^i$  is the position of vehicle  $i$  along lane  $\ell$  at time-step  $j$ .

- this scheme is free of numerical errors provided that

$$\Delta t = \frac{1}{w\kappa} \quad (11)$$

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# Quantizing $\Phi$

- Recall

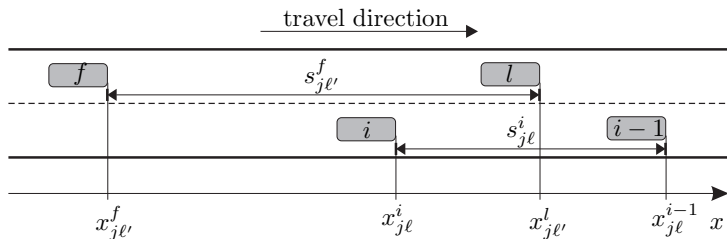
$$\gamma_{\ell'} = \min\left\{1, \frac{\mu_{\ell'}}{(1 - \sum_{\ell \neq \ell'} \pi_{\ell' \ell} \Delta t) \lambda_{\ell'} + \sum_{\ell \neq \ell'} \Delta x L_{\ell \ell'}}\right\} \quad (12)$$

- taking the limit  $\Delta t, \Delta x \rightarrow 0$

$$\Phi_{\ell \ell'} = \min\left\{1, \frac{\mu_{\ell'}}{\lambda_{\ell'}}\right\} \frac{\pi_{\ell \ell'} \lambda_{\ell}}{u} \quad (13)$$

- this continuum process needs to be quantized into vehicle units.

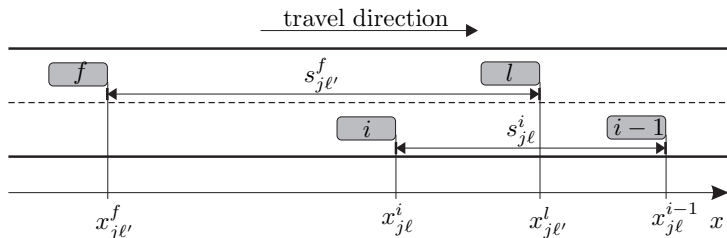
# Quantizing $\Phi$



- let  $p_{\ell\ell'}^{ij}$  be the probability of vehicle  $i$  in lane  $\ell$  of switching to lane  $\ell'$  during time-step  $j$ .

$$\begin{aligned}
 p_{\ell\ell'}^{ij} &= \int_{t=t_j}^{t_{j+1}} \int_{x=x_{j\ell}^i}^{x_{j\ell}^{i-1}} \Phi(k_\ell(t, x), k_{\ell'}(t, x)) dt dx \\
 &\approx \Phi(k_\ell(t_j, x_{j\ell}^i), k_{\ell'}(t_j, x_{j\ell}^i)) \Delta t s_{j\ell}^i
 \end{aligned} \tag{14}$$

# Quantizing $\Phi$



- Densities and spacings are related by

$$k_{\ell}(t_j, x_{j\ell}^i) = 1/s_{j\ell}^i \quad (15)$$

$$k_{\ell'}(t_j, x_{j\ell}^i) = 1/s_{j\ell}^f \quad (16)$$

where  $f$  gives  $i$ 's new follower in target lane  $\ell'$

- lane changes  $\sim \text{Bernoulli}(p_{\ell\ell'}^{ij})$

# Outline

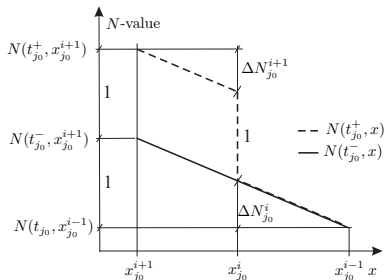
## 1 Background

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- **The microscopic moving bottleneck model**
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## Definition

Let  $N(t, x)$  be the vehicle number at  $(t, x)$  on the target lane under consideration, and

$$\Delta N_j^i = N(t_j, x_j^i) - N(t_j, x_j^{i-1})$$

the difference in  $N$  between vehicle  $i$  and its leader  $i - 1$

$$\Delta N_j^i = \begin{cases} 1, & i \text{ in "equilibrium"}; \\ < 1, & \text{o/w.} \end{cases}$$

- through line A-B pass  $h\kappa$  vehicles, thus

$$\Delta N_{j_0}^{i+1} = \Delta N_{j_0+1}^{i+1} = h\kappa \quad (17)$$

- from the figure

$$h = s_{j_0+1}^{i+1} \frac{w}{v_{j_0+1}^i + w} \quad (18)$$

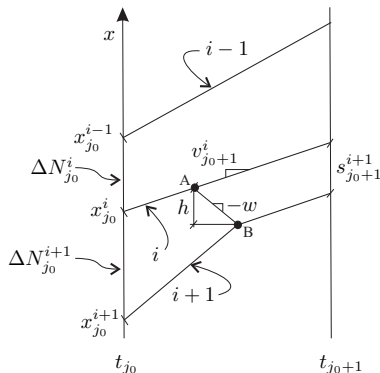
and therefore

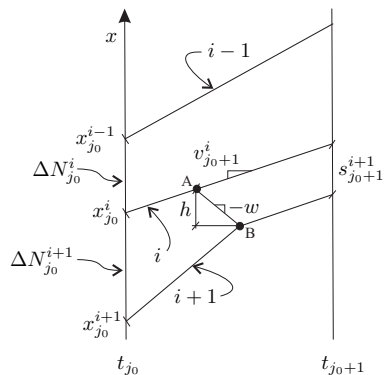
$$s_{j_0+1}^{i+1} = \frac{\Delta N_{j_0+1}^{i+1}}{K(v_{j_0+1}^i)} \quad (19)$$

where

$$K(v) = \frac{w\kappa}{v + w} \quad (20)$$

gives the density in congestion associated with speed  $v$ .





- also from the figure:

$$x_{j_0+1}^{i+1} = x_{j_0}^i + v_{j_0+1}^i \Delta t - s_{j_0+1}^{i+1} \quad (21)$$

Hence,

$$x_{j+1}^{i+1} = x_j^i + v_{j+1}^i \Delta t - \frac{\Delta N_{j+1}^{i+1}}{K(v_{j+1}^i)} \quad (22)$$

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# The continuum model

- Conservation law in Lagrangian coordinates

$$\frac{\partial s}{\partial t} + \frac{\partial v}{\partial N} = 0 \quad (23)$$

- for non-equilibrium drivers we propose

$$\frac{\partial v}{\partial N} = -\varepsilon \quad (24)$$

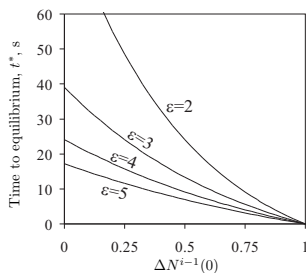
$\Rightarrow \varepsilon$  can be interpreted as the speed difference the follower is willing to accept with respect to its leader in order to attain an equilibrium spacing.

- we obtain:

$$\frac{d}{dt} \Delta N^{i+1}(t) = \varepsilon K(v^i(t)) \quad (25)$$

# The continuum model

## Example



bottleneck with constant acceleration rate  
 $\beta > 0$

$$\Delta N^{i+1}(t) = \Delta N^{i+1}(0) + \frac{\varepsilon w \kappa}{\beta} \ln \left[ 1 + \frac{\beta t}{v^i(0) + w} \right]$$

$\Rightarrow$  the time to convergence to equilibrium,  
 $t^*$ , equals

$$t^* = \frac{v^i(0) + w}{\beta} \left( \exp \left[ \beta \frac{1 - \Delta N^{i+1}(0)}{\varepsilon w \kappa} \right] - 1 \right)$$

## The discrete model I

- once  $v_{j+1}^{i+1}$  is known the follower spacing obeys

$$s_{j+1}^{i+1} = s_j^{i+1} + (v_{j+1}^i - v_{j+1}^{i+1})\Delta t. \quad (26)$$

- for  $j > j_0$ : differences in vehicle number and spacings are related by (19), ie

$$\Delta N_j^{i+1} = s_j^{i+1} K(v_j^i) \quad \text{with} \quad j > j_0. \quad (27)$$

- Combining (26)-(27) gives

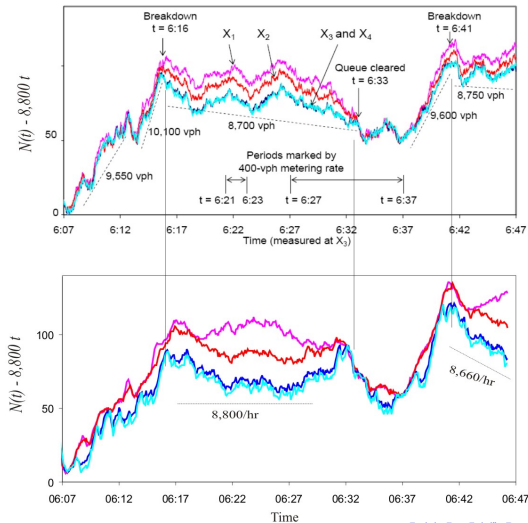
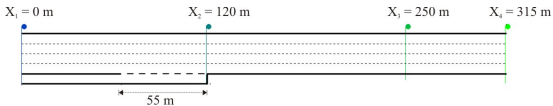
$$\Delta N_{j+1}^{i+1} = \left[ \frac{\Delta N_j^{i+1}}{K(v_j^i)} + (v_{j+1}^i - \tilde{v}_{j+1}^{i+1})\Delta t \right] K(v_{j+1}^i) \quad (28)$$

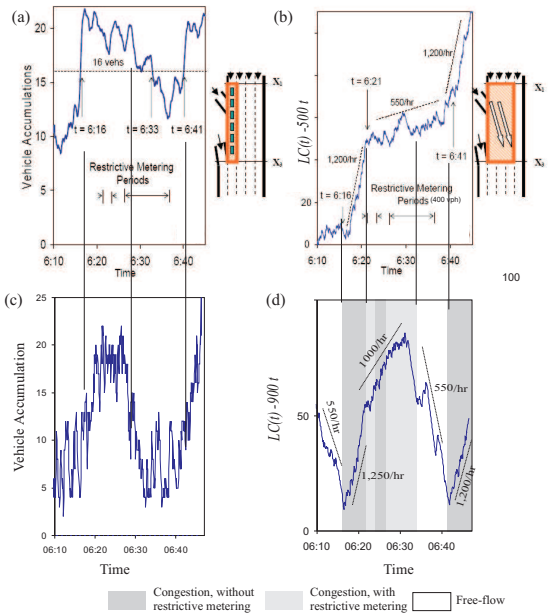
where we have introduced the follower's desired speed,  $\tilde{v}_{j+1}^{i+1}$ , since its actual speed  $v_{j+1}^{i+1}$  is not known at this stage.

# The discrete model II

- we propose

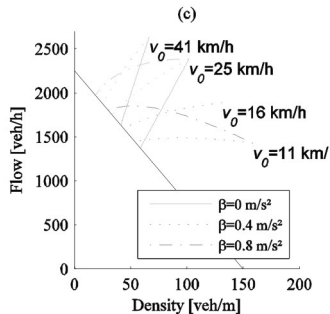
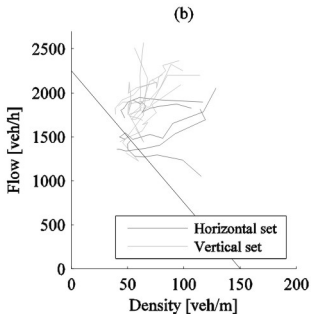
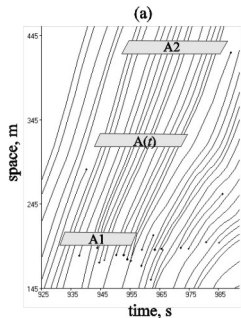
$$\tilde{v}_{j+1}^{i+1} = \underbrace{v_{j+1}^i(1 - \Delta N_j^{i+1}) + v_j^i \Delta N_j^{i+1}}_{\text{MB-speed}} - \varepsilon. \quad (29)$$



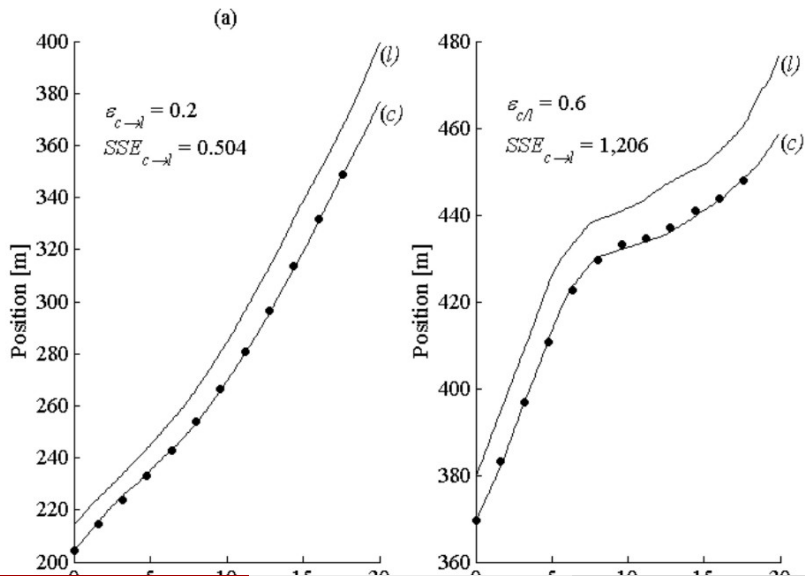


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# Relaxation paths



## Using trajectories



# Conclusions

- A complete multilane microscopic model with only two additional parameters:
  - $\tau$ : mean time to execute a lane change
  - $\varepsilon$ : speed-gap
- good agreement with field data
- continuum expression for  $\Phi$  can be used in many applications
- analytical solutions for KW with merges
- variational theory with LCs