

Part 2 - Modeling traffic flow

- engineering models
- microscopic models
- kinetic models
- macroscopic models

D. Helbing, A. Hennecke, and V. Shvetsov, Micro- and macro-simulation of freeway traffic. *Math. Computer Modelling* **35** (2002).

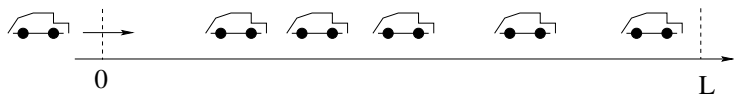
N. Bellomo, M. Delitala, V. Coscia, On the mathematical theory of vehicular traffic flow I. Fluid dynamic and kinetic modelling. *Math. Models Appl. Sci.* **12** (2002).

M. Garavello and B. Piccoli, *Traffic Flow on Networks. Conservation Laws Models*. AIMS Series on Applied Mathematics, Springfield, Mo., 2006.

A delay model (T. Friesz et al., 1993)

$X(t)$ = number of cars on a road at time t

If a new car enters at time t , it will exit at time $t + D(X(t))$



$D(X)$ = delay = total time needed to travel along the road

depends only on the total number of cars at the time of entrance

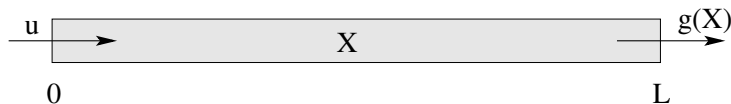
An ODE model

(D. Merchant and G. Nemhauser, 1978)

$X(t)$ = total number of cars on a road at time t

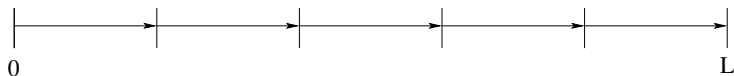
$u(t)$ = incoming flux $g(X(t))$ = outgoing flux

$$\dot{X}(t) = u(t) - g(X(t)) \quad \text{conservation equation}$$



L = length of road, $\rho \approx \frac{X}{L}$ = density of cars

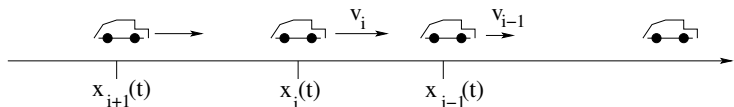
$$g(X) = \rho v(\rho) = \frac{X}{L} \cdot v\left(\frac{X}{L}\right)$$



Models favored by engineers:

- simple to use, do not require knowledge of PDEs (or even ODEs)
- easy to compute, also on a large network of roads
- become accurate when the road is partitioned into short subintervals

Microscopic models



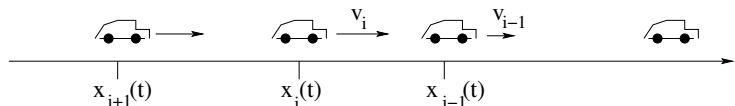
$x_i(t)$ = position of the i -th car

$v_i(t)$ = velocity of the i -th car

$i = 1, \dots, N$

Goal: describe the position and velocity of each car,
writing a large system of ODEs

Car following models



Acceleration of i -th car depends on:

- its speed: v_i
- speed of car in front: v_{i-1}
- distance from car in front: $x_{i-1} - x_i$

$$\begin{cases} \dot{x}_i = v_i \\ \dot{v}_i = a(v_i, v_{i-1}, x_{i-1} - x_i) \end{cases} \quad i = 1, \dots, N$$

Microscopic intelligent driver model (Helbing & al., 2002)

i -th driver $\left\{ \begin{array}{l} \text{accelerates, up to the maximum speed } \bar{v} \\ \text{decelerates, to keep a safe distance from the car in front} \end{array} \right.$

\bar{v} = maximum speed allowed on the road $v_i \in [0, \bar{v}]$

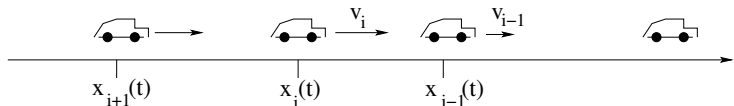
a = maximum acceleration

$$\dot{v}_i = a \cdot \left[1 - \left(\frac{v_i}{\bar{v}} \right)^\delta \right] - a \cdot \left(\frac{s^*(v_i, \Delta v_i)}{s_i} \right)^2$$

$s_i = x_{i-1} - x_i =$ actual gap from vehicle in front

$s_i^* =$ desired gap

Desired gap from the vehicle in front



$$s_i^* = \sigma_0 + \sigma_1 \sqrt{\frac{v_i}{\bar{v}}} + T v_i + \frac{v_i \Delta v_i}{2\sqrt{a} b} = \text{desired gap}$$

$\Delta v_i = v_i - v_{i-1} =$ speed difference with car in front

$\sigma_0 =$ jam distance (bumper to bumper)

$\sigma_1 =$ velocity adjustment of jam distance

$T =$ safe time headway

$b =$ comfortable deceleration

Equilibrium traffic

Assume: all cars have the same speed, constant in time.

Choose $\sigma_0 = \sigma_1 = 0$, $\delta = 1$

$$\dot{v}_i = a \cdot \left[1 - \frac{v_i}{\bar{v}} \right] - a \cdot \left(\frac{s^*(v_i, \Delta v_i)}{s_i} \right)^2 = 0$$

Equilibrium gap from vehicle in front

$$s_e(v) = s^*(v, 0) \cdot \left[1 - \frac{v_i}{\bar{v}} \right]^{-1/2}$$

Equilibrium velocity:
$$v_e(s) = \frac{s^2}{2\bar{v}T^2} \left(-1 + \sqrt{\frac{4T^2\bar{v}^2}{s^2}} \right)$$

$$\implies v_e = V_e(\rho) \quad \rho \approx s^{-1} = \text{macroscopic density}$$

Statistical (kinetic) description

$f = f(t, x, V)$ statistical distribution of position and velocity of vehicles

$f(t, x, V) dx dV$ = number of vehicles which at time t
are in the phase domain $[x, x + dx] \times [V, V + dV]$

local density: $\rho(t, x) = \int_0^\infty f(t, x, V) dV$

average velocity: $v(t, x) = \frac{1}{\rho(t, x)} \int_0^\infty V \cdot f(t, x, V) dV$

Evolution of the distribution function

$$\frac{\partial f}{\partial t} + V \frac{\partial f}{\partial x} + a(t, x) \frac{\partial f}{\partial V} = Q[f, \rho]$$

$a(t, x)$ = acceleration (may depend on the entire distribution f)

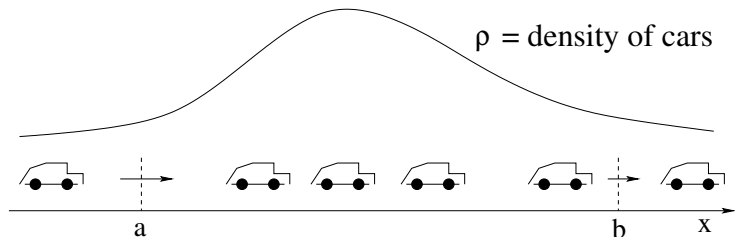
$Q(f, \rho)$ models a trend to equilibrium (as for BGK model in kinetic theory)

$$Q = c_r(\rho) \cdot \left(f_e(V, \rho) - f(t, x, V) \right)$$

c_r = relaxation rate

A conservation law model

(M. Lighthill and G. Witham, 1955)



$t = \text{time}$, $x = \text{space variable along road}$, $\rho = \rho(t, x) = \text{density of cars}$

flux: = [number of cars crossing the point x per unit time]

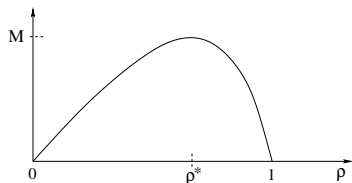
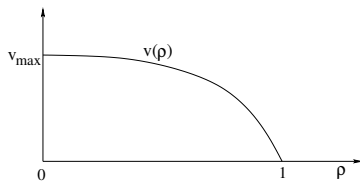
$$= [\text{density}] \times [\text{velocity}] = \rho \cdot v \quad v = V(\rho)$$

$$\rho_t + [\rho V(\rho)]_x = 0$$

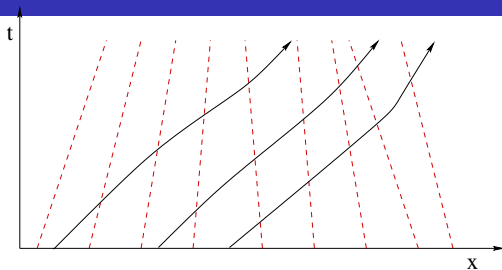
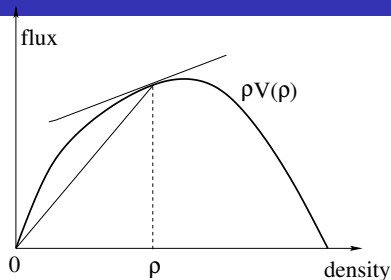
Flux function

Assume: $\rho \mapsto \rho V(\rho)$ is concave

$$V'(\rho) < 0, \quad 2V'(\rho) + \rho V''(\rho) < 0$$



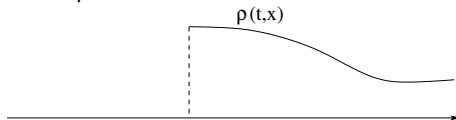
Characteristics vs. car trajectories



$$[\rho V(\rho)]' = V(\rho) + \rho V'(\rho) < V(\rho)$$

characteristic speed < speed of cars

Weak solutions can have upward shocks



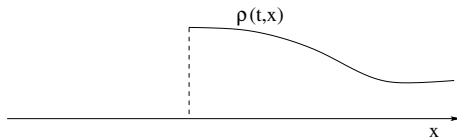
Adding a viscosity ?

$$\rho_t + [\rho V(\rho)]_x = 0 \quad (= \varepsilon \rho_{xx})$$

$$\rho_t + \left[\rho \left(V(\rho) - \varepsilon \frac{\rho_x}{\rho} \right) \right]_x = 0$$

effective velocity of cars: $v = V(\rho) - \varepsilon \frac{\rho_x}{\rho}$

can be negative, at the beginning of a queue



Second order models

$v = V(\rho) \implies$ velocity is instantly adjusted to the density

Models with acceleration

$$\begin{cases} \rho_t + (\rho v)_x = 0 \\ v_t + v v_x = a(\rho, v, \rho_x) \end{cases} \quad a = \text{acceleration}$$

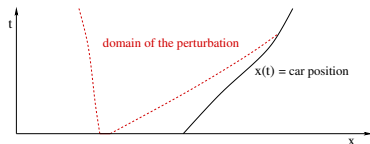
$$\begin{cases} \rho_t + (\rho v)_x = 0 \\ v_t + v v_x = \frac{1}{\tau}(V(\rho) - v) - \frac{p'(\rho)}{\rho} \rho_x \end{cases} \quad (\text{Payne - Witham, 1971})$$

$$[\text{relaxation}] + [\text{pressure term}] \quad p = \rho^\gamma, \quad \gamma > 0$$

$$\begin{cases} \rho_t + (\rho v)_x = 0 \\ v_t + v v_x + \frac{p'(\rho)}{\rho} \rho_x = \frac{1}{\tau}(V(\rho) - v) \end{cases}$$

$$\begin{pmatrix} \rho_t \\ v_t \end{pmatrix} + \begin{pmatrix} v & \rho \\ p'(\rho)/\rho & v \end{pmatrix} \begin{pmatrix} \rho_x \\ v_x \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

eigenvalues = characteristic speeds: $v \pm \sqrt{p'(\rho)}$



Wrong predictions: • negative speeds

• perturbations travel faster than the speed of cars

Idea: replace the partial derivative of the pressure $\partial_x p$ with the *convective derivative* $(\partial_t + v\partial_x)p$

$$\begin{cases} \partial_t \rho + \partial_x(\rho v) = 0 \\ \partial_t(v + p(\rho)) + v\partial_x(v + p(\rho)) = 0 \end{cases} \quad (\text{Aw - Rascle})$$

$$\begin{pmatrix} \rho_t \\ v_t \end{pmatrix} + \begin{pmatrix} v & \rho \\ 0 & v - \rho p'(\rho) \end{pmatrix} \begin{pmatrix} \rho_x \\ v_x \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

strictly hyperbolic for $\rho > 0$, positive speed: $v + p(\rho) \geq 0$

eigenvalues: $\lambda_1 = v - \rho p'(\rho), \quad \lambda_2 = v$

Properties of the Aw-Rascle model

- system is strictly hyperbolic (away from vacuum)
- the density ρ and the velocity v remain bounded and non-negative
- characteristic speeds (= eigenvalues) are smaller than car speed
 \implies drivers are not influenced by what happens behind them.
- maximum speed of cars on an empty road depends on initial data

An improved model (R. M. Colombo, 2002)

$$\text{Aw - Rascle:} \quad \begin{cases} \partial_t \rho + \partial_x(v\rho) = 0 \\ \partial_t q + \partial_x(vq) = 0 \end{cases}$$

$$q = v\rho + \rho p(\rho) = \text{"momentum"}$$

$$\text{Colombo:} \quad \begin{cases} \partial_t \rho + \partial_x(v\rho) = 0 \\ \partial_t q + \partial_x(v(q - q_{\max})) = 0 \end{cases}$$

$$v = \left(\frac{1}{\rho} - \frac{1}{\rho_{\max}} \right) q$$

ρ_{\max} = maximum density

q_{\max} = "maximum momentum"

\implies velocity can vanish only when $\rho = \rho_{\max}$,
and remains uniformly bounded

Concluding remarks

Number of vehicles on a road \ll number of molecules in a gas

- microscopic models (solving an ODE for each car) are within computational reach
- kinetic models and macroscopic models are realistic on longer stretches of road, for densities away from vacuum
- optimization problems, dependence of solution on parameters, are better understood by studying macroscopic models
- Simple ODE models, delay models are popular among engineers. Scalar conservation laws are OK. Kinetic models, second order models, are a hard sell.