# Weighted Monte Carlo method applied to acceleration oriented traffic flow model

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## Introduction

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### Main macro characteristics of the Vehicular Traffic Flow (VTF) are:

- 1. flux, flow Q [auto / time]
- 2. density *C* [auto / length]
- 3. mean velocity *V* [km/hour]

### Approaches to modelling: deterministic and probabilistic (stochastic).

- In the bases of the <u>deterministic</u> approach lies a functional relation between some of the main VTF characteristics, such as, e.g. velocity and the distance between the cars in the VTF.
- In <u>stochastic</u> models, oh the other hand, the VTF is considered as a probabilistic process.

#### All VTF models can be subdivided into three groups:

- <u>Analog models.</u> VTF is likened to a physical flow: fluid dynamic flow (**macroscopic** models) and gas dynamic flow (**mesoscopic** or **kinetic** models).
- **Microscopic** (<u>cellular automata</u>, <u>car following model</u>) models. These models are based on the assumption that there exists a relation between the movement of the leading and the following cars.

#### **VTF MODELS**

#### 1) Microscopic models (cellular automata, car following model)

This type of models regards the VTF as a system of interacting particles. Moreover, these models use Newton equations of motion, written individually for each particle in the system.



#### 2) Macroscopic models (<u>fluid dynamic</u>)

Within this type of models VTF is regarded as a one-dimensional flow of compressible fluid. We assume that.

**1.** The flow conservation low is valid for VTF (this condition is stated through the continuity equation; the on- and off-ramps are taken into account)

$$\frac{\partial c(x;t)}{\partial t} + \frac{\partial J(x;t)}{\partial x} = \sum_{i=1}^{J_{\text{in}}} \alpha_i (x - x_i;t) - \sum_{j=1}^{J_{\text{out}}} \beta_j (x - x_j;t)$$

2. There exists a one-to-one relation

between the flux and the density of the VTF

or

between the mean velocity and the density of the VTF

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### 3) Kinetic

Within the kinetic theory VTF is considered as a gas, i.e. a system of interacting particles; every particle in this system corresponds to a certain vehicle. These models are called mesoscopic due to the fact that they use information of the single vehicle behaviour as input and produce results of the whole traffic flow.

### The most obvious differences of VTF from gas flow are

- VTF is organized and mostly one dimensional;
- based on a deterministic set of rules;
- depends (more or less) on individual drivers.

### **Boltzmann kinetic equation:**

$$\left[\frac{\partial f}{\partial t} + \frac{p}{m} \cdot \nabla_r + F \cdot \nabla_p\right] f(r, p; t) = \left(\frac{\partial f}{\partial t}\right)_{\text{coll}} \text{ equation describes the}$$

equation describes the probability distribution.

 $\left(\frac{\partial f}{\partial t}\right)_{coll}$  this item is responsible for particle interactions in the system. We assume, that the state of a vehicle is determined by a number of characteristics. So, **interaction** of 2 vehicles means the **event**, which leads to any change in the state of these vehicles.

#### Acceleration oriented VTF model (K.T. Waldeer)

By analogy with gas kinetic transport theory, the state of a car is determined by a space coordinate and velocity in most mesoscopic VTF models. In this type of models the acceleration is a jump process in the velocity variable.

In the model, suggested by K.T. Waldeer, the acceleration variable is regarded as one of the phase coordinates, describing the state of a car. This modification leads to the discontinuous change of the acceleration variable as a result of the interaction in the system (not velocity variable, as it happens in most of kinetic models).

For a spatial homogeneous traffic flow the Boltzmann-like equation (BE) describing VTF has the following form

$$\begin{split} \frac{\partial f}{\partial t} + a \frac{\partial f}{\partial v} &= \int\limits_{\bar{v}, \bar{a}, a'} \left( \Sigma(a \mid v, a', \bar{v}, \bar{a}) f(a', v, t) - \right. \\ &\left. - \Sigma(a' \mid v, a, \bar{v}, \bar{a}) f(a, v, t) \right) f(\bar{a}, \bar{v}, t) d\bar{v} \, d\bar{a} \, da' \, . \end{split}$$

 $\Sigma$  is weighted interaction rate function (which determines the type of interaction in the system), f(v, a, t) is a one-particle probability density, describing the state of a car.

#### **Boundary condition:**

- 1) There are no cars with negative velocities (i.e. there are no cars with negative acceleration among the cars with V=0);
- 2) There is a maximum velocity of a VTF, which can not be exceeded (i.e. there are no cars with positive acceleration among the cars with  $V=V_{max}$ ).

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### **Initial velocity density without any acceleration: Numerical Solving Procedure (K.T.Waldeer)**

$$f(v, a, t = 0) = \frac{1}{\sqrt{2\pi\sigma_0}} e^{\frac{(v-V_0)^2}{2\sigma_0^2}} \delta(a)$$

We solve the equation with the help of the following  $f(v, a, t + \Delta t) = (1 - D\Delta t + J\Delta t)f(v, a, t) + O(\Delta t)$ splitting the movement and the interaction process  $= (1 - D\Delta t)(1 + J\Delta t)f(v, a, t) + O(\Delta t)$ 

in 2 parts executed consecutively in a time interval  $\Delta t$ :

Here  $\mathcal{J}$  is a nonlinear operator on the right side of BE, describing the discontinuous stochastic acceleration change,  $\mathcal{D} = a\partial/\partial v$  is a velocity drift operator on the left side of BE. Splitting process

The transition of a state probability density *f* to a new state at  $t+\Delta t$  is made in 2 steps:

- 1) First, in a given time interval  $\Delta t$  density f is changed in the acceleration variable resulting in a new, intermediate probability density  $f_I(v, a, t) = (1 + \mathcal{J}\Delta t)f(v, a, t)$
- 2) Then this intermediate probability density  $f_I$  is changed in velocity due to the drift operator  $\mathcal{D} = a\partial/\partial v$  into a new state probability density at  $t + \Delta t$

#### Simulation process:

- 1) Choose the integration time step  $\Delta t$
- 2) For each of *i*=1...*N* cars do
  - a) <u>Choose a leading car j</u>
  - b) Calculate a new acceleration for car number i  $a_i^*$
- 3) For each of i=1...N cars calculate new velocities  $(v_i + a_i^* \Delta t) \rightarrow v_i$
- 4) The next time step.

# Integral equation of the second kind

Density f(a, v, t) satisfies the Boltzmann type equation:

$$\frac{\partial f}{\partial t} + a \frac{\partial f}{\partial v} = \int_{\bar{v}, \bar{a}, a'} \left[ \Sigma(a|v, a', \bar{v}, \bar{a}) f(a', v, t) - \Sigma(a'|v, a, \bar{v}, \bar{a}) f(a, v, t) \right] f(\bar{a}, \bar{v}, t) \, \mathrm{d}\bar{a} \, \mathrm{d}\bar{v} \, \mathrm{d}a'$$

$$\tag{1}$$

We obtain an integral equation of the second kind equivalent to (1) to eliminate the parameter  $\Delta t$  from the simulation procedure.

Let us introduce the p.d.f. of N-particles system  $P(A, V, t) = P(a_1, v_1, \ldots, a_N, v_N, t)$ :

$$\begin{aligned} \frac{\partial P}{\partial t} + A \frac{\partial P}{\partial V} + \nu(A, V) P(A, V, t) &= J_N(A, V, t), \text{ here} \\ \nu(A, V) &= \frac{1}{N-1} \sum_{i \neq j} \int \Sigma(a_i \to a_i'' | v_i, v_j, a_j) \, \mathrm{d}a_i'' = \sum_{i \neq j} \frac{\nu(a_i, v_i, v_j, a_j)}{N-1} = \sum_{\pi} \frac{\nu(\pi)}{N-1}. \\ J_N(A, V, t) &= \int F(A' \to A | V) P(A', V, t) \, \mathrm{d}A' = \\ &= \frac{1}{N-1} \sum_{i \neq j} \int \left[ \Sigma(a_i' \to a_i | v_i, a_j, v_j) \prod_{m \neq i, m=1}^N \delta(a_m' - a_m) \right] P(A', V, t) \, \mathrm{d}A'. \end{aligned}$$

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It is reasonable to use collision density  $\Phi(A, V, t) = \nu(A, V)P(A, V, t)$  rather than function P for simulating the Markov chain. Moreover, it is even more convenient to use function  $\Psi(A, V, t)$ :

$$\Phi(A, V, t) = \int_{0}^{t} \int \Psi(A, V', t') K_{t}(t' \to t | A, V') \, \mathrm{d}t' K_{V}(V' \to V | A, t - t') \, \mathrm{d}V',$$

In this case the free term is nonzero only for t = 0:

$$\Psi(A, V, t) = \delta(t)P_0(A, V) +$$

 $\iiint_{0}^{t} \Psi(A',V',t')K_{t}(t'\to t|A',V')\,\mathrm{d}t'K_{V}(V'\to V|A',t-t')\,\mathrm{d}V'K_{A}(A'\to A|V)\,\mathrm{d}A',$ 

the transition densities are

$$K_t(t' \to t | A', V') = \chi(t' < t)\nu(A', V' + A'(t - t')) \exp\left\{-\int_{t'}^t \nu(A', V' + A'(\tau - t')) \, \mathrm{d}\tau\right\},\$$
  
$$K_V(V' \to V | A', t - t') = \delta(V - V' - A'(t - t')), \quad K_A(A' \to A | V) = \frac{F(A' \to A | V)}{\nu(A', V)}.$$

The transitions from the state (A', V', t') to the state (A, V, t) is performed as follows

$$(A', V', t') \to (A', V', t) \to (A', V, t) \xrightarrow{\pi} (A, V, t).$$

# Stratification according to the pair number

We add the index  $\pi$  of the pair of cars participating in the current interaction in the system to the phase space. Function  $\Psi$  can be represented in the form:

$$\Psi(A,V,t) = \sum_{\pi} F(\pi, A, V, t).$$

We obtain the integral equation for the function F in the modified phase space  $(Z,t)=(\pi,A,V,t)$ :

$$F(Z,t) = \delta(t)P_0(A,V)\delta(\pi_0) + \int_0^t \int F(Z',t')K(Z',t' \to Z,t) \,\mathrm{d}Z' \,\mathrm{d}t'.$$

Here  $\pi_0$  is an arbitrary pair number, and the kernel K is the factorization of the transition densities:

$$K(Z', t' \to Z, t) = K_t(t' \to t | A', V') K_V(V' \to V | A', t - t') K_\pi(\pi) K_a(a'_i \to a_i | \pi, V).$$

# Estimation of the functionals

Usually the following linear functionals are of interest

$$I_h(T) = \iint h(v, a) f(T, v, a) \, \mathrm{d}v \, \mathrm{d}a.$$

One can show that [Mikhailov, Rogasinsky]

$$I_h(T) = \iint_0^T H(A, V + A(T - t')) \exp\left\{-\int_{t'}^T \nu(A, V + A(\tau - t') \,\mathrm{d}\tau\right\} F(\pi, A, V, t') \mathrm{d}Z \mathrm{d}t',$$

here  $H(V, A) = \frac{1}{N} \sum_{i=1}^{N} h(v_i, a_i).$ 

For estimation of the latter integral we can use standard collision estimator  $\xi$  or absorbtion one  $\eta$ :

$$\xi = \sum_{k=0}^{S} Q_k \tilde{H}(A_k, V_k, t_k, T),$$
$$\eta = Q_S \frac{\tilde{H}(A_k, V_k, t_k, T)}{p(A_k, V_k, t_k, T)}.$$

# Majorant frequency method

Let  $\nu_{max} \ge \nu(\varpi)$  for any  $\varpi = (i, j)$  and  $\nu^* = \sum_{i \ne j} \nu_{max}/(N-1) = N \cdot \nu_{max}$ . Then

$$\frac{\partial P}{\partial t} + A \frac{\partial P}{\partial V} + \nu^* P(A, V, t) = J_N^*(A, V, t) = \nu^* \int K^*(A' \to A | V) P(A', V, t) \, \mathrm{d}A',$$

$$K^* = \sum_{\pi = (i,j)} K^*_{\pi} \left[ \left( 1 - \frac{\nu'(\pi)}{\nu_{max}} \right) \delta(a'_i - a_i) + \frac{\nu'(\pi)}{\nu_{max}} \frac{\Sigma(a'_i \to a_i | v_i, v_j, a'_j)}{\nu'(\pi)} \right] \prod_{m \neq i} \delta(\ldots).$$

 $K_{\pi}^* = [N(N-1)]^{-1}$  is equiprobability distribution of the pair number  $\pi$  of cars participating in interaction. Time distribution transforms to the exponential one:

$$K_t^*(t' \to t) = \chi(t' < t)\nu^* \exp\left\{-\nu^*(t - t')\right\},\,$$

For the function  $F^*(Z, t)$  we obtain an integral equation with the kernel

$$K(Z', t' \to Z, t) = K_t^*(t' \to t) K_V(V' \to V | A', t - t') K_\pi^* K_a(a'_i \to a_i | \pi = (i, j), V).$$
$$I_h(T) = \iint_0^T H(A, V + A(T - t')) \exp\left\{-\nu^*(T - t')\right\} F^*(\pi, A, V, t') \, \mathrm{d}Z \, \mathrm{d}t'.$$

# Model Cases Compared with Analytical Solutions

We consider all the cars to have bounded velocities  $0 \le v_i \le V_{max}$  and a single value acceleration  $\pm a_0$ .

We tested two interaction types to compare with the analytical solutions.

#### 1) Maxwell interaction

$$\Sigma(a'_i \to a_i | v_i, a_j, v_j) = \frac{1}{\mathcal{T}} \begin{cases} \delta(a - a_0), & v_i \leq v_j, \\ \delta(a + a_0), & v_j < v_i. \end{cases}$$

In this case  $\nu(\pi) = const = 1/\mathcal{T}, \ \nu^* = N/\mathcal{T}$ . Under these conditions, kinetic equation can be solved analytically in stochastic equilibrium, resulting in

$$f(v,a) = \frac{\pi}{4\sqrt{3}\sigma_v} \cosh^{-2}\left\{\frac{\pi}{2\sqrt{3}}\frac{(v-V)}{\sigma_v}\right\} \frac{\delta(a-a_0) + \delta(a+a_0)}{2},$$

with the mean V and the variance  $\sigma_v^2 = (\pi \mathcal{T} a_0)^2/3$ . We used the following parameters in the computer simulation with N = 1000 stochastic cars and M = 1000 runs:  $\sigma_0^2 = 0.1 \text{ m}^2/\text{s}^2$ , V = 20 m/s,  $a_0 = 0.3 \text{ m/s}^2$ , T = 18 s,  $\mathcal{T} = 2 \text{ s}$ ,  $\sigma_v = 1.088 \text{ m/s}$ .

#### 2) Hard sphere interaction

$$\Sigma(a'_i \to a_i | v_i, a_j, v_j) = r_0 | v_j - v_i | \begin{cases} \delta(a - a_0), & v_i \le v_j, \\ \delta(a + a_0), & v_j < v_i. \end{cases}$$

In this case  $\nu(\pi) = r_0 |v_i - v_j|$ ,  $\nu_{max} = r_0 V_{max}$ ,  $\nu^* = N V_{max} r_0$ . The solution in stochastic equilibrium is given by

$$f(v,a) = \frac{1}{\sqrt{2\pi}\sigma_v} \exp\left\{-\frac{(v-V)^2}{2\sigma_v^2}\right\} \frac{\delta(a-a_0) + \delta(a+a_0)}{2},$$

with the mean V and the variance  $\sigma_v^2 = a_0/r_0$ . We used the following parameters in the computer simulation with N = 1000 stochastic cars and M = 1000 runs:  $\sigma_0^2 = 0.1 \text{ m}^2/\text{s}^2$ , V = 20 m/s,  $a_0 = 0.3 \text{ m/s}^2$ , T = 62 s,  $r_0 = 0.25 \text{ m}^{-1}$ ,  $\sigma_v = 1.095 \text{ m/s}$ .

# References

[1] K.T. Waldeer. The direct simulation Monte Carlo method applied to a Boltzmann-like vehicular traffic flow model // Comp. Phys. Commun. 2003. V. 156, N 1, pp. 1–12.



Figure 1: Estimation of f(v) for the Maxwell interaction, T = 18 s: 1 - exact solution, 2 - estimation, 3 - confidence interval  $\pm 3\sigma$ .



Figure 2: Estimation of f(v) for the hard sphere interaction, T = 62 s: 1 - exact solution, 2 - estimation, 3 - confidence interval  $\pm 3\sigma$ .