

Mathematical modeling by differential equations

Case study: Traffic flow

Stefan Doboszczak, Virginia Forstall

University of Maryland

M3C

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Outline

- 1 Introduction
- 2 Case study: A simple model for traffic flow
- 3 Case study: Extension of traffic flow model
- 4 Numerical techniques
- 5 Criticism

Introduction



Introduction

- 1 State the problem.
- 2 Identify variables and effects.
- 3 Choose modeling method.
- 4 Formulate basic model.
- 5 Build on basic model.
- 6 Solve equations or run simulations.
- 7 Make conclusions and criticize model.

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- Types of possible models:
 - ▶ Cellular automata: Follow individual cars, give rules for motion
 - ▶ PDE: Ignore individual cars, consider average densities
- PDE approach: Formulate a model for the density $\rho(x, t)$, of cars per unit length, as a function of position x and time t , given the initial density $\rho(x, 0)$.
- Automata approach: Create a simulation of cars on a 1d lattice by specifying rules for their movement on the lattice.

Relevant variables / effects

Variables:

Relevant variables / effects

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- position, x
- time, t
- density, $\rho = \rho(x, t)$
- velocity, $v = v(x, t)$
- flux, $J = J(x, t)$

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Effects to account for:

- length of cars
- spacing between cars
- length of roadway
- number of lanes
- merging
- on/off ramps
- weather
- accidents/construction

Assumptions

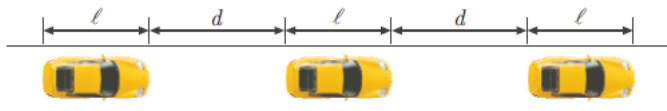
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- Assume motion is directed left-to-right on a one-lane roadway of infinite length. Is this reasonable?

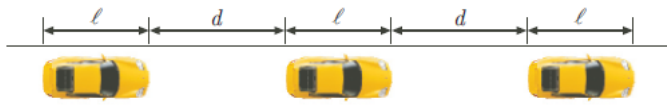
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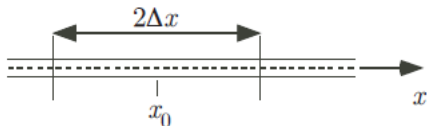
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- Ignore all other effects.

Uniform distribution model: density

- Consider the interval $I_{x_0} = \{x_0 - \Delta x < x < x_0 + \Delta x\}$ about the point x_0 .



To estimate $\rho(x_0, t_0)$, count the number of cars in the interval and divide by its length:

$$\rho(x_0, t_0) \approx \frac{1}{2\Delta x} (\# \text{ of cars in the interval } I_{x_0} \text{ at time } t = t_0) \quad (1)$$

- The assumption on Δx is that it is small enough so only cars in the vicinity matter but large enough so the interval contains many cars. The continuum view has $\Delta x \rightarrow 0$.

Uniform distribution model: density

- Recall the assumption is cars have length L and uniform spacing d . Using (1) we find that

$$\rho(x_0, t_0) = \lim_{\Delta x \rightarrow 0} \frac{1}{2\Delta x} \left(\frac{2\Delta x}{L+d} \right) = \frac{1}{L+d}$$

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- Note that since $0 \leq d < \infty$ and L is finite that there is a maximum density:

$$0 < \rho \leq \rho_{max} = \frac{1}{L}.$$

- For instance, if $L = 17\text{ft}$ and $d = 12\text{ft}$, then $\rho = 182 \frac{\text{cars}}{\text{mile}}$ and $\rho_{max} = 310.6 \frac{\text{cars}}{\text{mile}}$.

Uniform distribution model: flux

- Question: How many cars pass the point x_0 per unit time?
- To find the flux of cars through x_0 at time t_0 , count the number of cars passing x_0 during the time interval

$$I_{t_0} = \{t_0 - \Delta t < t < t_0 + \Delta t\}$$

and divide by the interval length. In other words,

$$J(x_0, t_0) \approx \frac{1}{2\Delta t} (\# \text{ of cars passing } x_0 \text{ during the time interval } I_{t_0}) \quad (2)$$

- The assumption on Δt is that it is small enough so only cars passing by near time t_0 are counted, but big enough so that many cars pass through. Again, in the continuum limit, $\Delta t \rightarrow 0$.

Uniform distribution model: flux

- Assume that cars move to the right with constant speed v .
- During time $2\Delta t$, cars move a distance equal to $2v\Delta t$, hence the number of cars passing a point x_0 during this time is $\rho \times \text{distance} = \frac{2v\Delta t}{L+d}$. Using (2), the flux about x_0 at time t_0 is

$$J(x_0, t_0) = \lim_{\Delta t \rightarrow 0} \frac{1}{2\Delta t} \left(\frac{2v\Delta t}{L+d} \right) = \frac{v}{L+d}.$$

- If for instance $L = 17\text{ft}$ and $v = 70\text{mph}$, then $J = 5435 \frac{\text{cars}}{\text{mile}}$.
- Note also that

$$J = \rho v.$$

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Balance law for density

- The previous model is extended to allow for changes in density. This is accomplished by using a **control volume** (CV) and accounting for all the ways the density can change in the volume.
- Assumptions: cars cannot enter through the top or bottom (no ramps), cars enter from the left and exit from the right (one way traffic).

Balance law for density

- Consider the CV consisting of the interval I_x during the time interval I_t . The idea is simple, the net change of cars in I_x during I_t is equal to the net flux of cars through the boundaries during this time. In other words:

$$\begin{aligned} & \{\# \text{ of cars in } I_x \text{ during time } t_0 + \Delta t\} - \{\# \text{ cars in } I_x \text{ during time } t_0 - \Delta t\} \\ &= \{\# \text{ of cars that enter through } x_0 - \Delta x \text{ during } I_t\} \\ & - \{\# \text{ of cars that leave through } x_0 + \Delta x \text{ during } I_t\} \end{aligned}$$

Balance law for density

- In terms of density and flux:

$$\begin{aligned}2\Delta x[\rho(x_0, t_0 + \Delta t) - \rho(x_0, t_0 - \Delta t)] \\ = 2\Delta t [J(x_0 - \Delta x, t_0) - J(x_0 + \Delta x, t_0)]\end{aligned}$$

- The goal is, as before, to take $\Delta x, \Delta t \rightarrow 0$ to deduce the continuum limit. This is accomplished by expanding ρ and J in Taylor series and considering the leading order terms.

Balance law for density

- By Taylor expanding we find that (about the point (x_0, t_0)):

$$\begin{aligned} & 2\Delta x[\rho + \Delta t \rho_t + \frac{1}{2}(\Delta t)^2 \rho_{tt} + O((\Delta t)^3) \\ & \quad - \rho - (-\Delta t \rho_t) - \frac{1}{2}(\Delta t)^2 \rho_{tt} + O((\Delta t)^3)] \\ &= 2\Delta t[J - \Delta x J_x + \frac{1}{2}(\Delta x)^2 J_{xx} + O((\Delta x)^3) \\ & \quad - J - \Delta x J_x - \frac{1}{2}(\Delta x)^2 J_{xx} + O((\Delta x)^3)] \end{aligned}$$

- This reduces to

$$2\Delta x[2\Delta t \rho_t + O((\Delta t)^3)] = 2\Delta t[-2\Delta x J_x + O((\Delta x)^3)]$$

or as $\Delta x, \Delta t \rightarrow 0$,

$$\frac{\partial \rho}{\partial t} = -\frac{\partial J}{\partial x}. \quad (3)$$

Balance law for density

- If the flux function J is known, then (3) tells us how ρ evolves. This is usually deduced empirically.
- Instead focus on v . Assume $J = \rho v$ as before, where $v = v(x, t)$ is the average velocity of the cars at (x, t) .
- The PDE model for traffic flow then reads:

$$\begin{cases} \frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x}(\rho v) & = 0 \\ \rho(x, 0) & = f(x) \end{cases}$$

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- PDE can now be solved analytically or numerically.
- BUT, the PDE model is an approximation! Discrete traffic flow has been modeled by continuous density.
- Low densities are therefore a problem. We don't expect the PDE to hold when there are few cars on the road.
- The velocity v is still needed. We need a **constitutive law** relating v to ρ . This can be obtained by examining the physics or determined from empirical evidence.

Determining velocity: empirical data

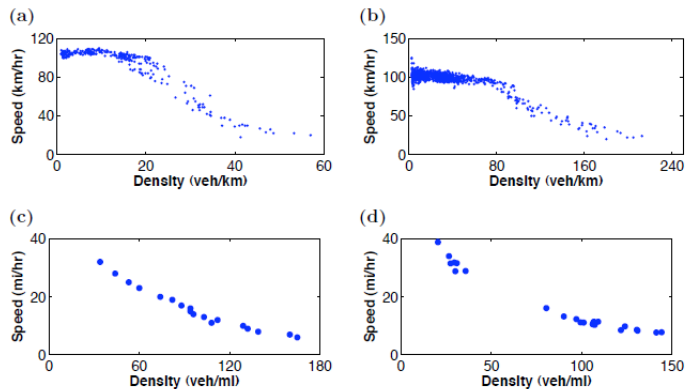


Figure 5.6 The velocity as a function of the density as measured for different roadways. Shown is (a) a highway near Toronto, (b) a freeway near Amsterdam, (c) the Lincoln Tunnel, and (d) the Merritt Parkway. Data for (a) and (b) are from Aerde and Rakha [1995], and (c) and (d) are from Greenberg [1959].

Determining velocity

- Assume the velocity is constant,

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Then

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- This perhaps is too unrealistic, except maybe low densities.

Determining velocity

- Assume a linear velocity,

$$v = v(\rho) = a - b\rho.$$

This is known as the Greenshields model. It is commonly written as

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

- From the Merritt Parkway and Lincoln Tunnel data, we can use least squares to find that $v_{max} = 36.821\text{mph}$ and $\rho_{max} = 166.4226 \frac{\text{cars}}{\text{mile}}$ are the constants of best fit for the linear velocity.

Determining velocity

- Using this form of v in our PDE we get that now

$$\rho_t + c(\rho)\rho_x = 0,$$

where $c(\rho) = v_{max} \left(1 - \frac{2\rho}{\rho_{max}}\right)$.

- Other more complicated velocity fits can be considered.

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Numerical Solution

Return to the model with a constant velocity on a finite length highway:

$$\frac{\partial \rho}{\partial t} + a \frac{\partial \rho}{\partial x} = 0 \quad \text{for } 0 < x < L, t > 0 \quad (4)$$

with an initial condition

$$\rho(x, 0) = \begin{cases} 1 & \text{if } x \leq 0 \\ 0 & \text{if } x > 0 \end{cases}$$

and a boundary condition

$$\rho(0, t) = g(t) .$$

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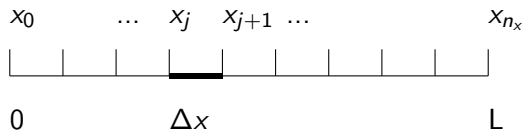
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Question?

Why is boundary specified at $x = 0$ and not at $x = L$?

Discretized problem

- Goal: Given $g(t)$ and a , numerically compute the density at time T .
- Basic Idea: Discretize the time interval $[0, T]$ into intervals of length Δt and the spatial interval $[0, l]$ into intervals of length Δx and use a finite difference to approximate the derivative.



Finite differences

Recall the definition of the derivative

$$\lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} \quad (5)$$

Let $x = x_j$ and $h = \Delta x$. Without the limit, we have forward difference of the first derivative

$$\frac{f(x_{j+1}) - f(x_j)}{\Delta x} . \quad (6)$$

The backward difference of the first derivative is

$$\frac{f(x_j) - f(x_{j-1})}{\Delta x} . \quad (7)$$

Finite differences

Let us return to our model,

$$\frac{\partial \rho}{\partial t} + a \frac{\partial \rho}{\partial x} = 0 \quad \text{for } 0 < x < L, t > 0 \quad (8)$$

Denote $\rho(x_j, t_n) = \rho_j^n$.

Finite differences

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Denote $\rho(x_j, t_n) = \rho_j^n$.

Use the forward difference to approximate the time derivative and the backward difference to approximate the spatial derivative:

$$\frac{\rho_j^{n+1} - \rho_j^n}{\Delta t} + a \frac{\rho_j^n - \rho_{j-1}^n}{\Delta x} = 0. \quad (9)$$

Thus we have approximations of the two derivatives, which will approach continuous equations in the limit as $\Delta x \rightarrow 0$ and $\Delta t \rightarrow 0$.

Finite differences

Thus our numerical scheme is

$$\rho_j^{n+1} = \rho_j^n + \frac{a\Delta t}{\Delta x}(\rho_j^n - \rho_{j-1}^n) \quad (10)$$

Example 1

- Cars move at $a = 30$ mph or $1/120$ miles per second.
- The highway is 3 miles long $L = 3$.
- Let's run the model for 3 minutes which means $T = 180$ s.
- $\Delta x = 0.1$ mile, $\Delta t = 0.1$ s
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What does this boundary condition mean in our real world scenario?

Example 2

The stoplight turns red at $t = 1$ minute and green again at $t = 2$ minutes. There is a constant stream of cars that want to go through the stoplight.

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$$g(t) = \begin{cases} 1 & \text{if } 0 < t < 60, t \geq 120 \\ 0 & \text{if } 60 \leq t < 120 \end{cases}$$

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Criticism

- We have ignored many effects mentioned earlier, for instance, intersections, weather, traffic lights, etc.
- Perhaps our guess on velocity was too restrictive. A form such as $v = F(\rho, \rho_x)$ may be better.
- For highway travel, merging and multilane effects become important.
- The models presented seem to work only for light to heavy traffic situations.

References

- [1] M. Holmes
Introduction to the Foundations of Applied Mathematics.
Springer (2009)