# Artificial Intelligence to calibrate traffic models

**ERASMUS+ Mobility** 

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USP | EESC

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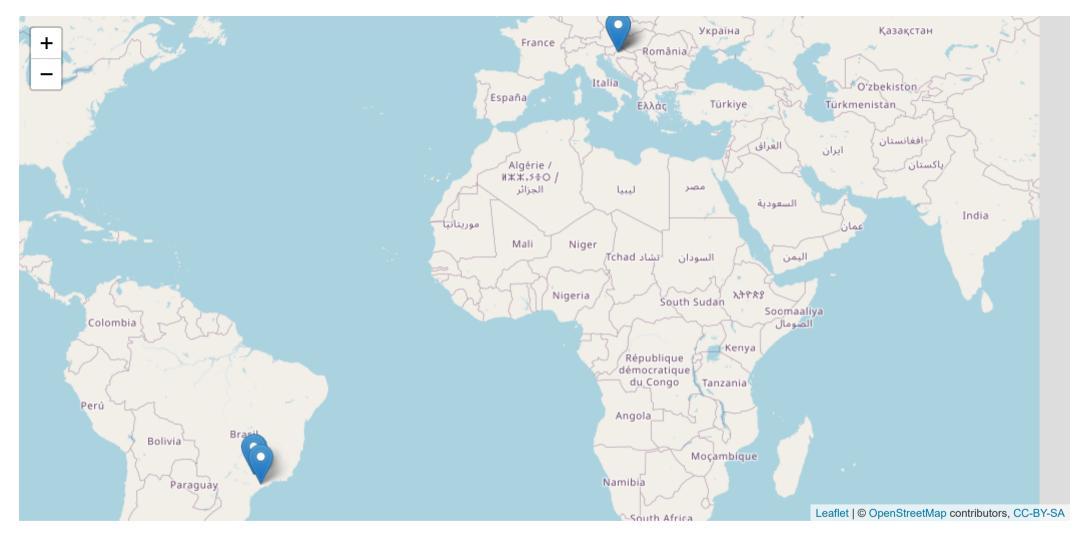


University of Zagreb
Faculty of Transport and Traffic Sciences





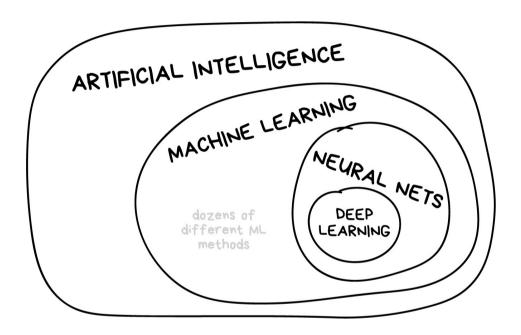
## Virtual Mobility



website: **USP** 

## **Get Started**

## Concepts



- |AI| Artificial Intelligence: a whole knowledge field
- |ML| Machine Learning: is a part of AI, but not the only one
- |NN| Neural Networks: one of the machine learning types
- |DP| Deep Learning: a modern method of building, training and using NN (a new architecture)

Image credit: vask3

## Methods & Algorithms

Image credit: vask3

## Genetic Algorithm [GA]

Concept

Algorithm

Elements

Operators

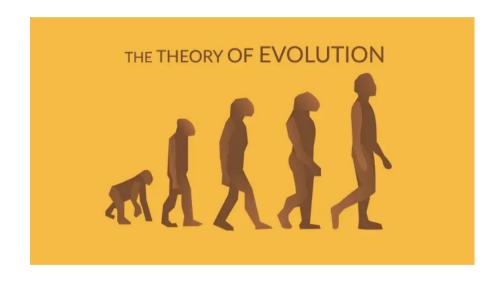


Image credit: EarthHow

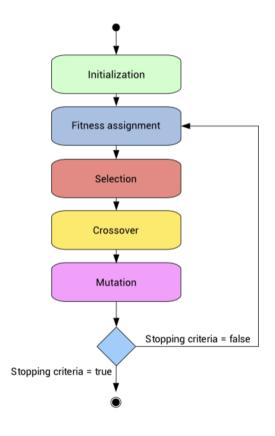
GAs are stochastic search algorithms inspired by principles of biological and natural selection.

-- *John H. Holland* (1984)

GAs simulate the evolution of living organisms, where the fittest individuals dominate over the weaker ones, by mimicking the biological mechanisms of evolution, such as selection, crossover and mutation.

-- [STRUCCA, 2021]

Concept Algorithm Elements Operators



• Free for any kind of model:

$$y = f(lpha_n, X_m)$$
  $model = f(parameters, variables)$   $min(Error) = g(model, data)$ 

• Dataset:

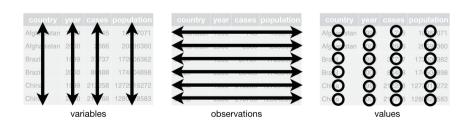
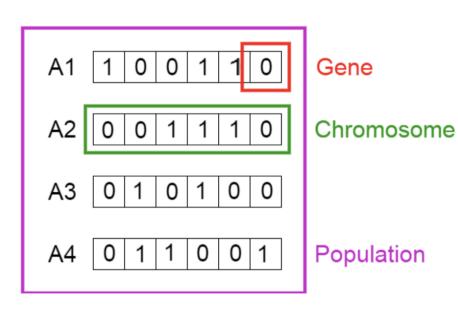


Image credit: R4DS

Concept Algorithm Elements Operators



$$chromosome = \left[\underbrace{\underbrace{1110101}_{gen\ 1}}\underbrace{\underbrace{1001}_{gen\ 2}}...\underbrace{\underbrace{10101}_{gen\ N}}\right]$$

$$chromosome = \begin{bmatrix} \underbrace{11}_{gen\ 1} \underbrace{32}_{gen\ 2} \underbrace{5}_{gen\ 2} \dots \underbrace{10}_{gen\ N} \end{bmatrix}$$

Concept

Algorithm

Elements

**Operators** 

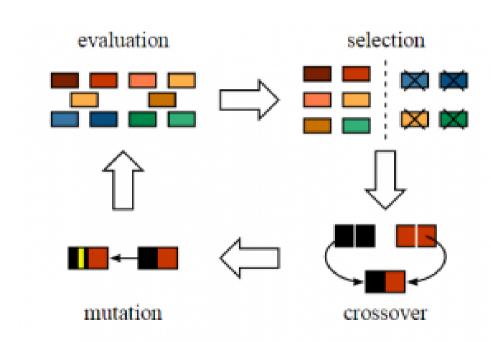


Image credit: PapersWithCode

#### Parameters to set:

- Evaluation:
  'fitness function' rank of the individuals
  Hollander & Liu, 2008
- **Selection/Eletism:** preserve the *n*-th best solutions
- **Crossover**: mix the solutions
- Mutation: change part of a solution
- **Predation**:

  "diversity", eliminate *k* worst solutions and generate new ones randomly

## Aplications of GA

## **Examples in Transport Engineering**

Research Team

Brazilian road transport



Prof. J.R. Setti [USP-EESC]



Prof. A.L. Cunha [USP-EESC]



Prof. J.E. Bessa Jr. [UFMG]

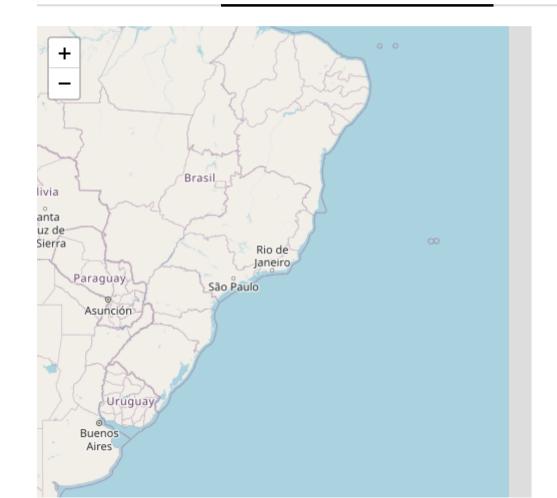


**Gustavo Riente** McTrans Center, UFL

### Examples in Transport Engineering

Research Team

Brazilian road transport





- ~80% of the roads are Two-lane Highways
- ~60% of Brazilian cargo is transported by road
- Need for *Highway Capacity Manual* for Brazil

### Microscopic models calibration

#### PREMISSES:

Microscopic models are essential tools to the Transport engineer, however he/she needs specialized knowledge to specify how well the simulation is reproduced.

The use of simulation models without proper calibration results in planning and design errors.

#### **MODELS:**

#### **Driver behavior:**

- Car-following logic
- Lane change logic

#### **Vehicle performance:**

- Acceleration \ Deceleration
- Emergency stop

Details

GA

Result

Publication

Developed by Australian Road Research Board (ARRB), Transport Research Ltd, Australia.

- Simulation time step: 1.0s
- 7 fleet composition
- 18 different vehicles
- 18 driver profiles

- coded in Perl
- 25 parameters calibrated
- Vehicle performance:
  - used 7 vehicles in 3 classes
     (car, unit truck, articulated truck)
  - Weight
  - Lenght
  - Weight/Power ratio
- Driver behavior:
  - 7 drivers profile
  - Free Flow Speed
  - Speed factor reduction in auxiliary lanes

- Error drops from 9% to 6% using calibrated parameters
- Method to estimate traffic parameters from video camera (manually).

- Egami, C.Y. [2000] (Master)
- Egami, C.Y. [2006] (PhD thesis)

Details

GA

Result

Publication

developed by Midwest Research Institute (MRI), Federal Highway Administration (FHWA), in USA.

- Simulation time step: 1.0s
- 13 different vehicles
- 10 driver profiles

- coded in Perl and Lua
- 35 parameters calibrated
- Vehicle performance:
  - used 4 vehicles
  - Weight/Power ratio
- Driver behavior:
  - 10 drivers profile
  - Free Flow Speed
  - Probability distribution of overtaking
  - Car-following factor

- Error drops from 9.5% to 6% using calibrated parameters
- Method to estimate traffic parameters from video camera (manually/automatic).

- [2006] Automatic Calibration of Two-Lane Highway Traffic Simulation Models Using Genetic Algorithm.
- [2008] Mon-Ma, M.L. (PhD thesis)
- [2011] Derivation of ATS and PTS function for Two-lane, Rural Highways in Brazil.

Details GA Truck model Result

developed by **Michael Van Aerde** and **Dr. Hesham Rakha**, at *Virginia Tech Transportation Institute* (VTTI), in USA. Simulate any type of highway (two-lane, multilane and freeway).

- Simulation time step: 0.1s
- 7 different vehicles
- 10 driver profiles
- Small version is available for download: https://sites.google.com/a/vt.edu/hrakha/software

Details GA Truck model Result

- Coded in VBA-Excel and Lua
- 8 parameters calibrated
- Vehicle performance:
  - 4 different vehicles
  - Weight/Power ratio
- Fitness function:
  - MAER (Mean Absolute Error)

Details GA Truck model Result

- Collect data:
  - GPS position of trucks at 10Hz (0.1s) travelling 10km on road
  - o 62 trucks (5 light, 13 medium, 22 heavy, and 22 extra-heavy)
- Model:

$$F=m\cdot a$$
  $F_t=min(\eta\cdot 3600\cdot rac{P}{V};W_{ta}\cdot \mu)$   $R_t=R_a+R_r+R_g$   $R_a=c_1\cdot C_D\cdot C_h\cdot A\cdot V^2$   $R_r=C_r\cdot (c_2\cdot V+c_3)\cdot rac{W}{1000}$   $R_g=W\cdot i$ 

Details GA Truck model Result

- The GA was able to find parameters to represent the average performance of four different trucks
- The MAER for truck classes ranged from 4.0% to 6.5%
- The MAER for individual trucks was 2.2%

Details GA Truck model Driver model Result Publication

developed by *Federal Highway Administration* (FHWA), in USA. It is capable to simulate freeways, two-lane highways, street in an integrated system.

- Simulation time step: 1.0s
- 7 different vehicles
- 10 driver profiles

- 31 parameters in TSIS-CORSIM
- Models were calibrated separately
- Vehicle performance:
  - 4 different vehicles
  - Weight/Power ratio
- Driver behavior:
  - 10 drivers profile
  - Car-following logic
  - Lane-changing logic
- Fitness function:
  - MAER (Mean Absolute Error)

- Collect data:
  - GPS position of trucks at 10Hz (0.1s) travelling 10km on road
  - o 62 trucks (5 light, 13 medium, 22 heavy, and 22 extra-heavy)
- Model:

- Collect data:
  - Speed and Flow data on a 10km of multilane highway
  - Speed and Flow data on 16 segments of two-lane highways
- Model:

- The AG was able to calibrate parameter to four different trucks and 10 different drivers profile.
- The MAER for truck classes ranged from 4.5% to 7.5%
- The MAER for individual trucks was 2.2%
- The error ratio dropped by almost 50% using the parameter calibrated for driver behaviour
- The method reveals how to estimate percentage time spent following (PTSF) with a calibrated model

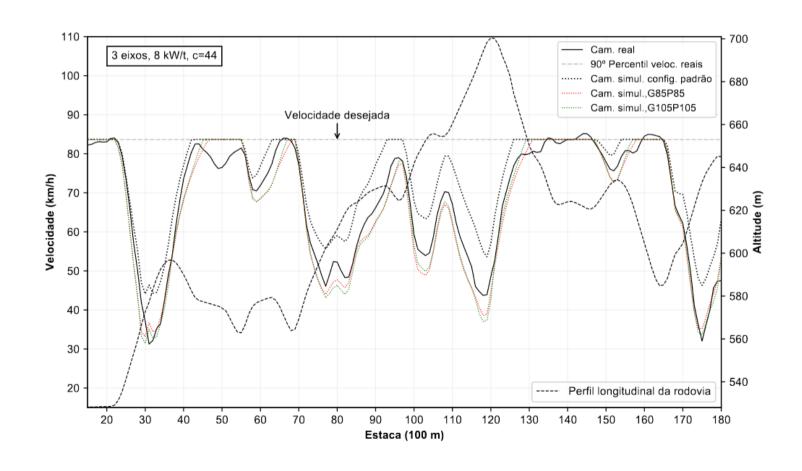
- [2007] Araújo, J.J. (PhD thesis)
- [2007] Cunha, A.L. (Master's)
- [2009] Genetic Algorithm for the calibration of vehicle performance models of microscopic traffic simulators.
- [2011] Truck equivalence factors for divided, multilane highways in Brazil.
- [2017] Evaluation of Models to Estimate Percent Time Spent Following on Two-Lane Highways.

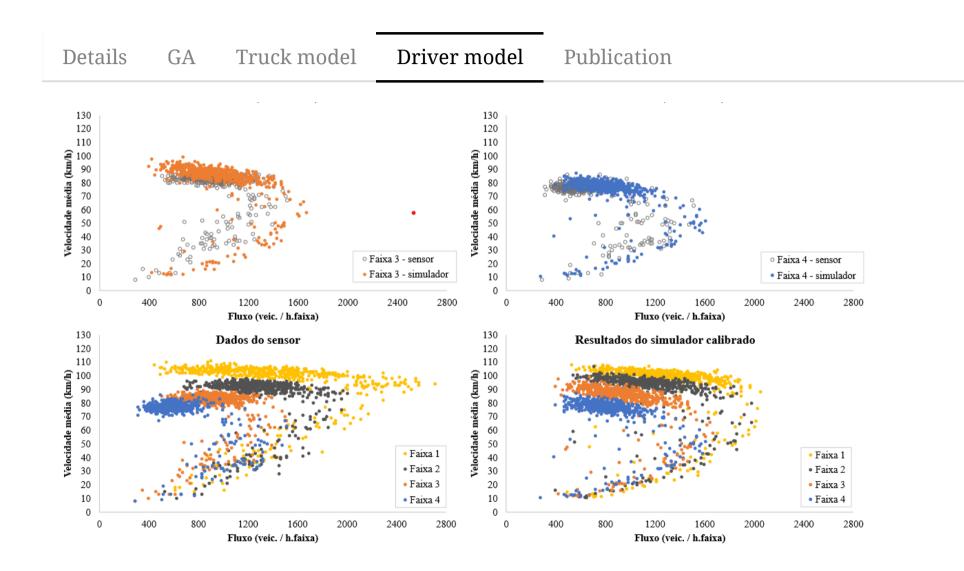
Details GA Truck model Driver model Publication

Developed by PTV Group in Karlsruhe, Germany. It is a microscopic multi-modal traffic flow simulation package.

- Simulation time step: 0.1s
- unlimited vehicles
- 10 driver profiles

- Coded in Python
- 30 parameters for each truck class
- 10 parameters for each driver profile
- Vehicle performance
  - used 4 truck classes
  - distribution of acceleration, power and weight
- Driver behavior
  - 10 drivers profile
  - Car-following logic
  - Lane-changing logic
- Fitness function
  - MAER (Mean Absolute Error)





- [2016] Calibration of microscopic traffic simulators using macroscopic measures
- [2018] Calibration of Vissim's truck acceleration model

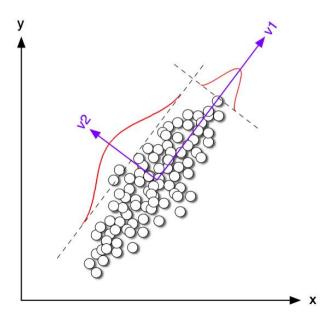
## Principal Component Analysis (PCA)

## Principal Component Analysis

#### Concept

Application

- Reducing dimensionality of data
- Preserve as much as possible the information contained in original data
- Emphasizes variation and highlights patterns in dataset



### Principal Component Analysis

Concept

Application

- [2016] A Model for Estimating Free-Flow Speed on Brazilian Expressways
- Models for estimation of FFS
  - HCM:

$$FFS = BFFS - f_1 - f_2 - \ldots - f_n$$

• HDM-4:

$$FFS = f(V_{drive}, V_{brake}, V_{curve})$$

• Proposed:

FFS = f(PostSpeed, NumLanes, HwyType, AdjLandUse, HorzAlign, AccessDen)

## Bayesian Inference

### Bayesian Inference

Concept

**Application** 

- Specify how one should update one's belief upon observing data
- A method of statistical inference used to update the probability for a hypothesis as more evidence or information become available.

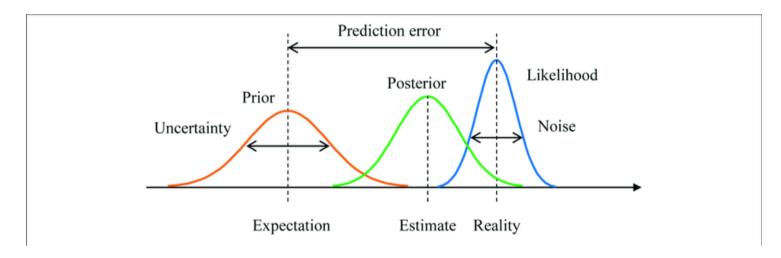


Image credit: Yanagisawa

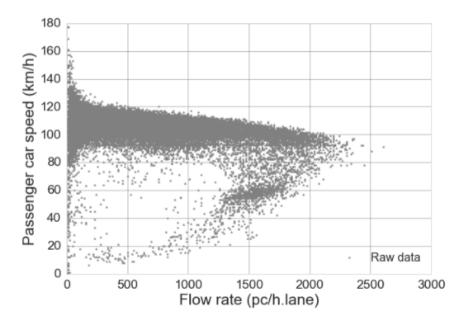
## Bayesian Inference

Concept

Application

- [2019] A calibration method structured on Bayesian Inference of the HCM speed-flow relationship for freeways and multilane highways and a temporal analysis of traffic behavior
- HCM Speed-flow model:

$$u=u_f-((u_f-rac{q_c}{k_c})\cdot (rac{q-bp}{q_c-bp})^lpha)\cdot (q\geq bp)$$



## Thanks!

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https://github.com/albnc