



ICTR 2023



11th INTERNATIONAL CONGRESS on TRANSPORTATION RESEARCH
Clean and Accessible to All Multimodal Transport
Heraklion, Crete, September 20th - 22nd 2023

Modelling the behaviour of automated vehicles when interacting with jaywalkers

Foteini Orfanou

Transportation Engineer, PhD Candidate

Together with:
Eleni Vlahogianni, George Yannis



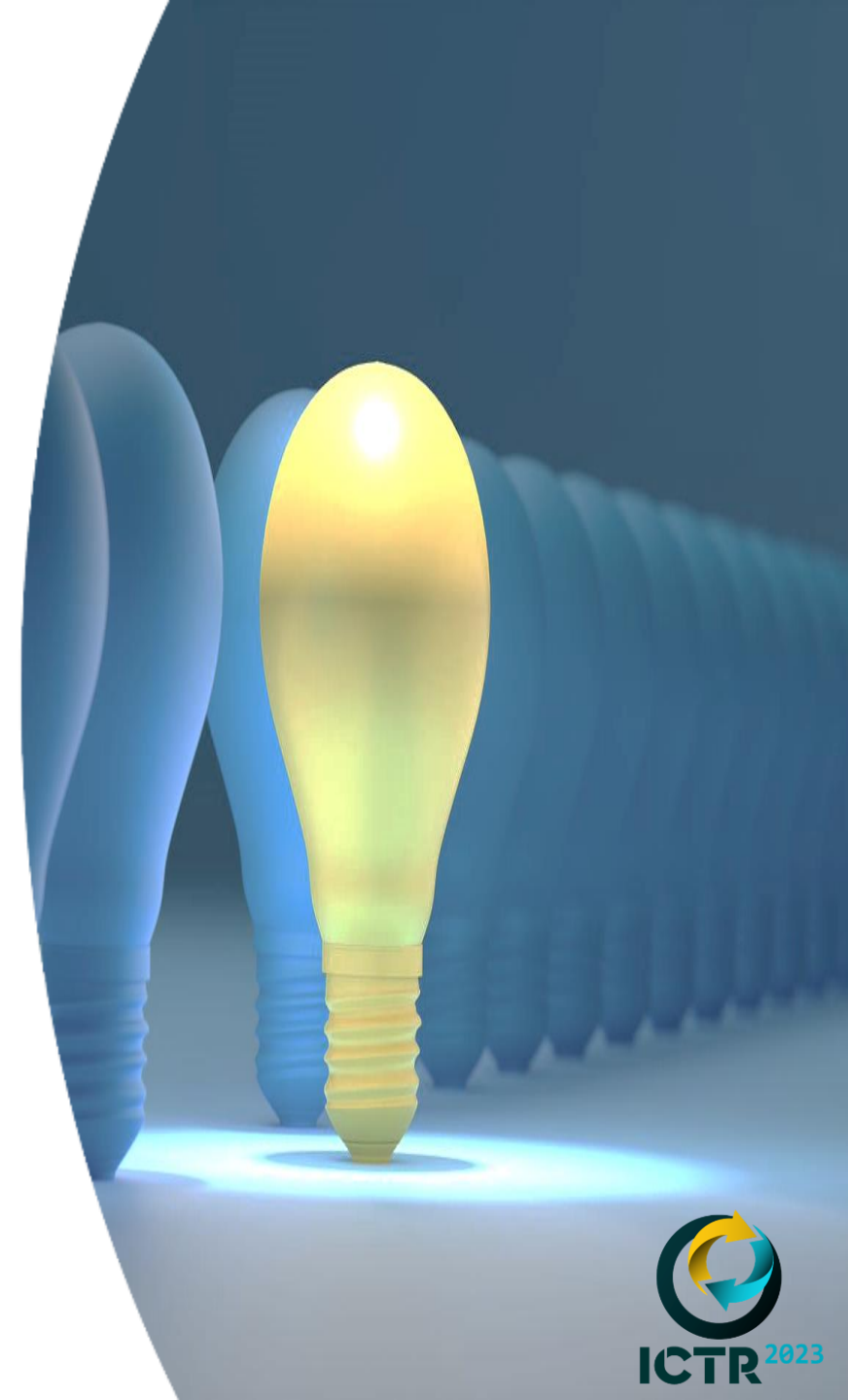
Introduction

- The advent of **automation** revolutionizes the transportation field
- Contribution towards **safer vehicles** and **safer people**
- **Safe** interaction between **VRUs** and **equipped vehicles** is a major road safety pillar
 - **Illegal crossing** is considered the riskiest
- Need to understand the **microscopic characteristics** governing these encounters



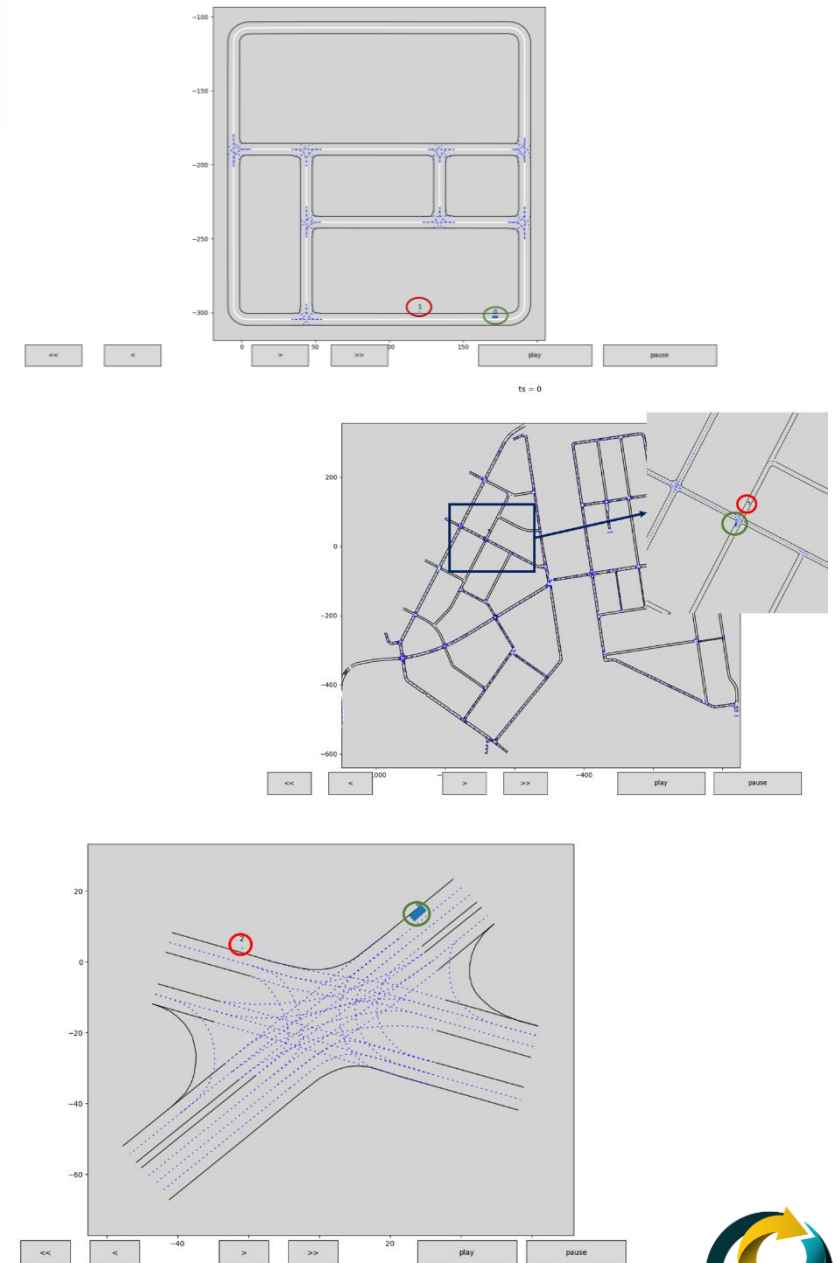
Objectives

- Identification of the most **critical and appropriate indicators** for describing the vehicle-pedestrian interaction
- Classify the interactions in different severity levels
- Deeper understanding of the AV-pedestrian interaction process dynamics based on **Inverse Reinforcement Learning** principles



The Experiment

- **Virtual Experiment** led by FZI Research Centre for Information Technology
- **Level 4 automated** vehicle
- **Human expert** immersed into the scene via a virtual reality (VR) headset
- Road crossing at a random location of the network
- Data collection in **3 different** phases and **networks**

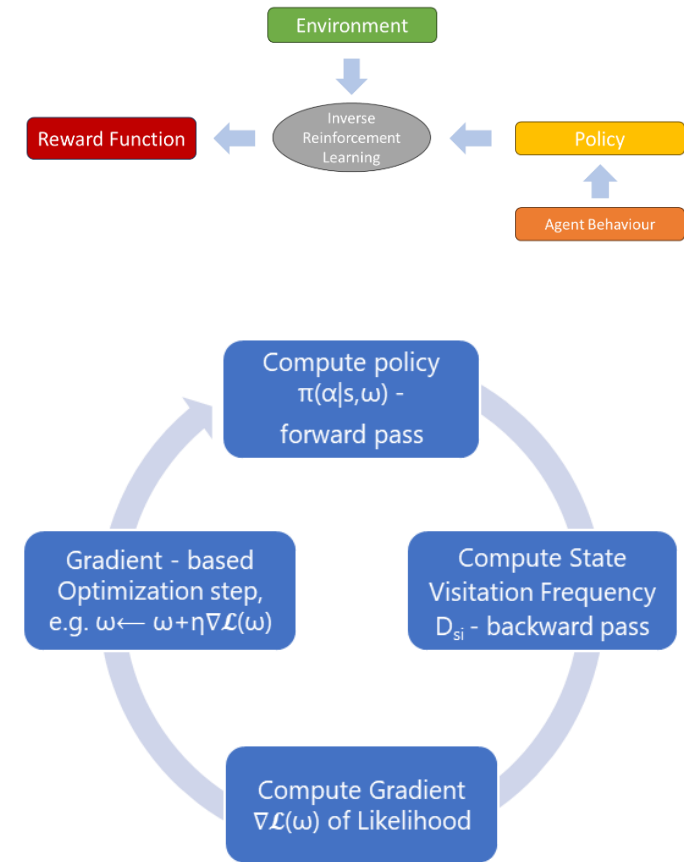
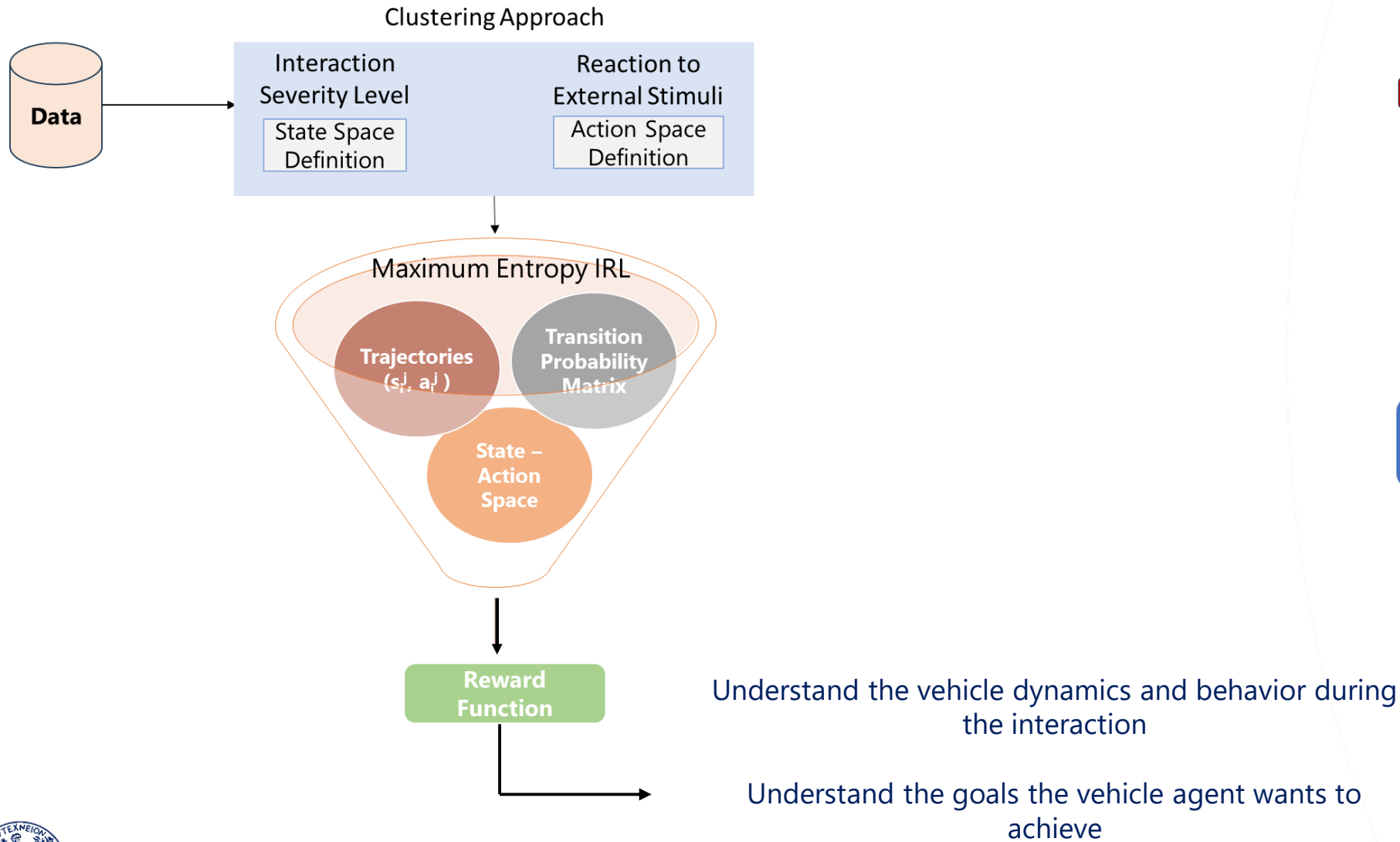


Data Collection

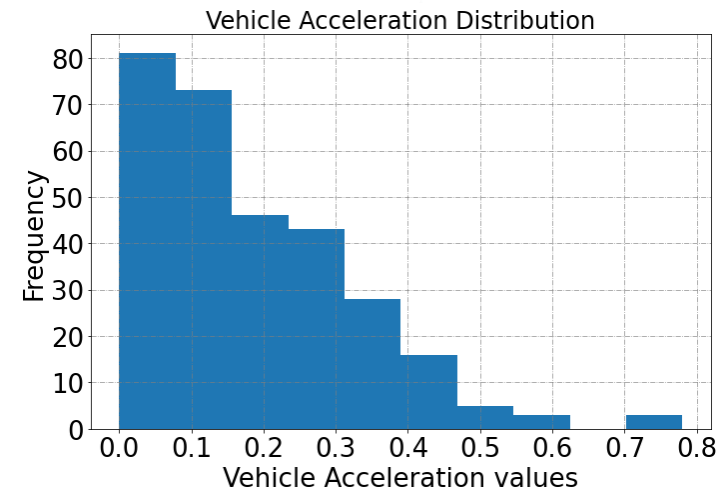
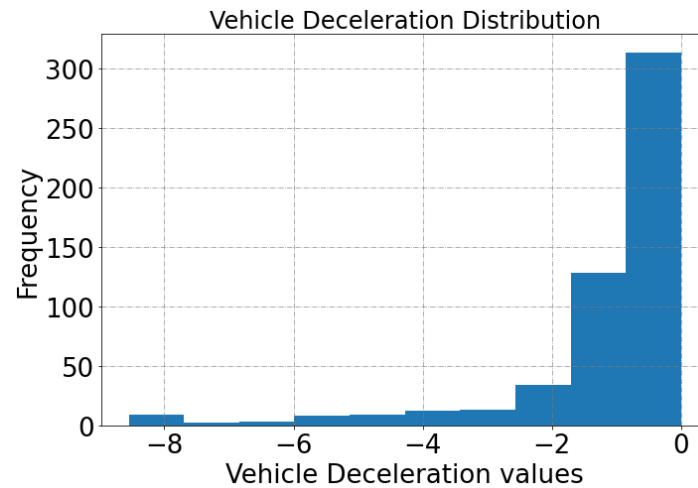
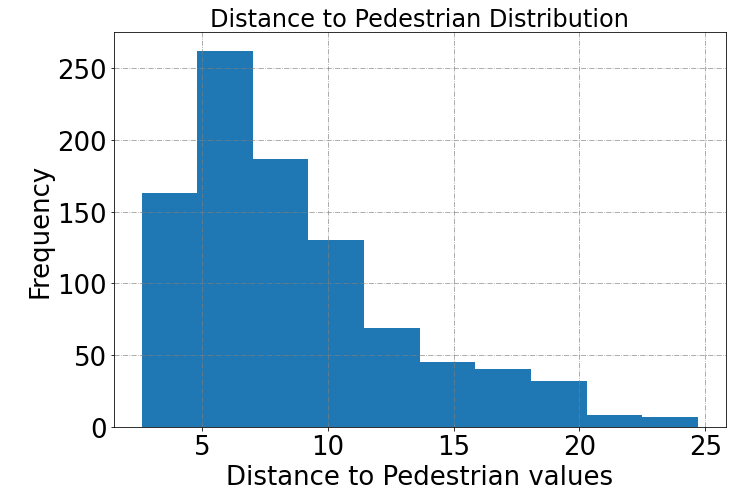
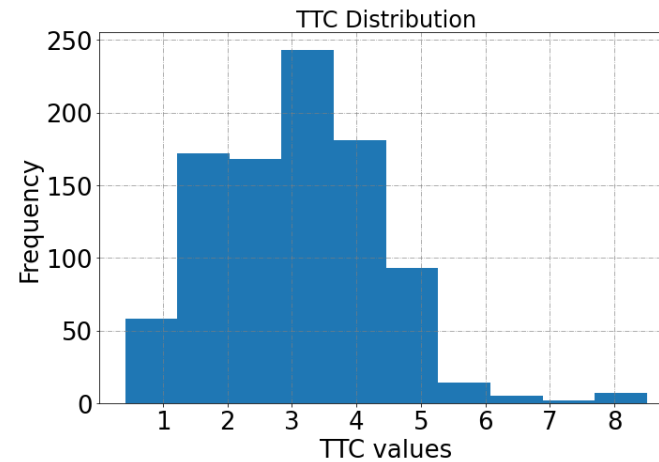
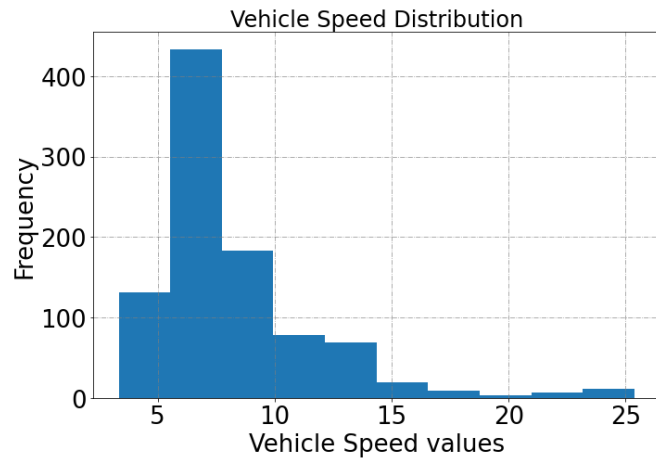
- **Kinematic characteristics** of the two agents
- **Spatial and temporal** distances
- **Phase 1:** 21 pair of trajectories
- **Phase 2:** 21 pair of trajectories
- **Phase 3:** 45 pair of trajectories
- **Final dataset:** 50 trajectories
 - Pedestrian passes first

Field	Description	Unit
Timeframe	data is collected every 100 ms	ms
Agent -type	"car" or "pedestrian"	-
x,y	the x and y position of the agent	m
vx, vy	speed values of the agent in the x and y dimension	m/s
psi_rad	the yaw angle of the agent	rad
length	length of the agent	m
width	width of the agent	m
ax, ay	acceleration/deceleration values of the agent in the x and y dimension	m/s ²
time-headway	the temporal distance of the agent from its preceding vehicle	s
gap	the spatial distance of the agent from its preceding vehicle	m
lateral_position	the distance of the central axis of the vehicle from the central axis of the lane	m
side_distance	the distance from the central axis of the agent from the side object	m
mode	this parameter describes whether automation mode is on or manual vehicle control is applied. Since automation can be activated only in case the agent-type is a car, this parameter takes the values "automated" and "simulated" if automated function is on and off respectively. For pedestrians, the only value for this parameter is "simulated".	

Methodological Overview

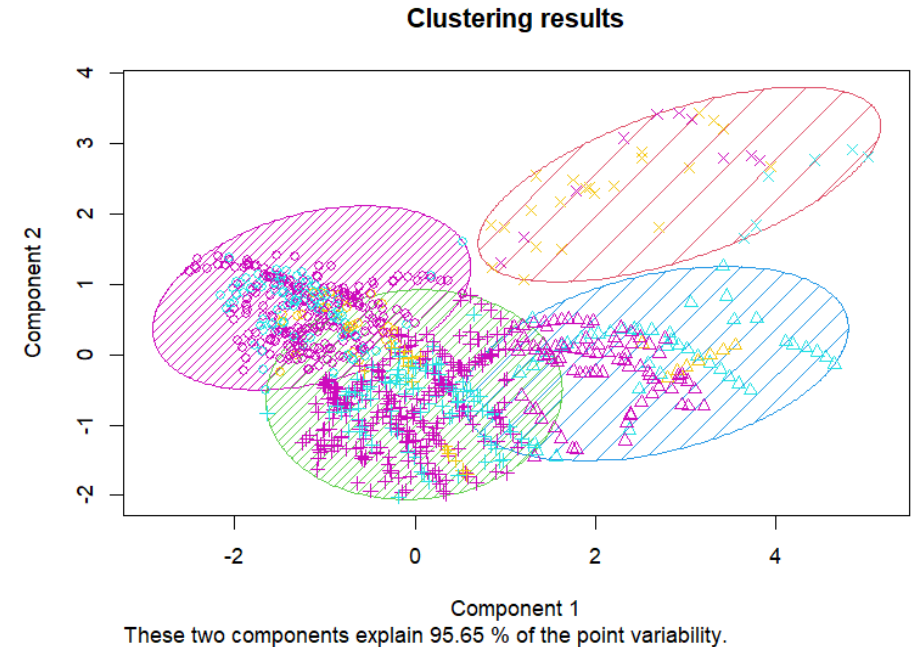


Descriptive Statistics



Vehicle State Space Definition

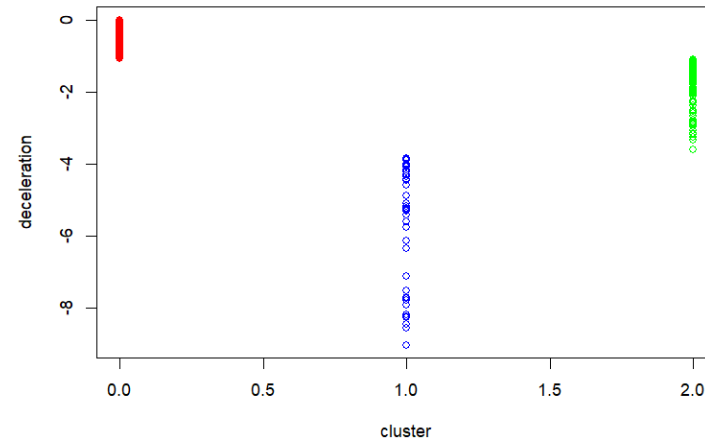
- Combination of **time-, distance – and speed - based** indicators
- **Data driven unsupervised learning**
 - K – means Clustering
- **Vehicle Speed, Gap and Time to Collision (TTC)**
- Definition of the **severity of the interaction** process
- Number of clusters=4
- Silhouette Coefficient = **0.462**



	Average Speed	Maximum Speed	Standard Deviation of Speed	Average Distance	Maximum Distance	Standard Deviation of Distance	Average TTC	Maximum TTC	Standard Deviation of TTC	Sate Characterization
Cluster 0	6.91	13.24	1.61	5.60	9.5	1.51	1.83	2.65	0.54	Unsafe
Cluster 1	12.45	18.14	1.78	16.17	24.7	3.01	4.01	4.95	0.54	Safe
Cluster 2	7.11	12.01	1.58	7.90	13.06	2.34	3.64	4.96	0.58	Safe
Cluster 3	19.82	25.37	3.78	14.56	22.69	3.55	2.15	3.17	0.41	Critical

Vehicle Action Space Definition

- Critical parameter to define the driver reaction to the external stimuli
 - Acceleration/Deceleration
- Based on k-means clustering results:
 - 1 levels of acceleration
 - 3 levels of deceleration (smooth, harsh, emergency)
- 5 actions
 - Do nothing
 - Acceleration
 - Smooth deceleration ($> -1.08\text{m/s}^2$)
 - Harsh deceleration ($< -3.6\text{m/s}^2$)
 - Emergency deceleration ($< -8.45\text{m/s}^2$)

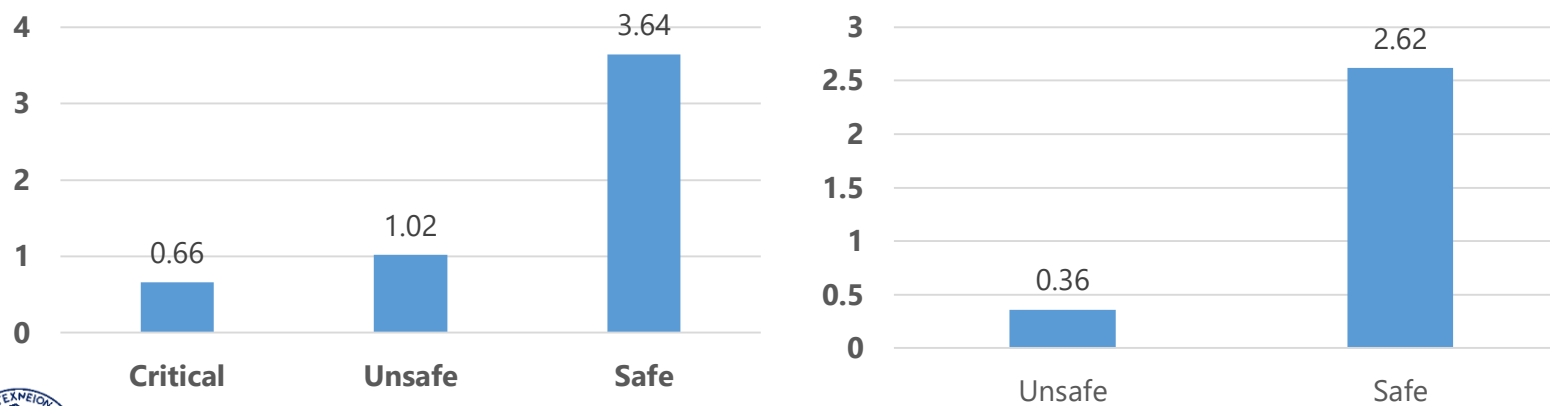


	Average Deceleration	Maximum Deceleration	Standard Deviation of Deceleration	Action Characteriza tion
Cluster 0	-0.46	-1.08	0.33	Smooth
Cluster 1	-1.73	-3.60	0.56	Harsh
Cluster 2	-5.66	-8.45	1.65	Emergency

Actions	Value
Cruising (no change in speed)	0
Acceleration	(0, 0.78]
Smooth deceleration	[-1.08, 0)
Harsh deceleration	[-3.6, -1.08)
Emergency deceleration	[-8.45, -3.6)

MaxEnt IRL Results

- As the risk level increases the reward decreases and the vehicle "driver" gets penalized
- The expert adopts the vehicle dynamics to be more frequent under the state "safe"
- The experts goal is to take actions for avoiding the "critical" and "unsafe" states
- The reward function can be used to learn but not copy the expert' policy



Conclusions

- Clustering approach allows easier and more reliable interpretation of the reward function
- Need for additional parameter for improving the vehicle state space definition
- As expected, safer states tend to give higher rewards to the vehicle “driver”
- The transition to unsafe and dangerous states penalizes the vehicle “driver”



Further Research

- Consideration of additional parameters for improving clustering results and vehicle state space definition
 - Better depict the collision proximity and severity
- Optimization of vehicle control strategy
 - Reinforcement Learning principles
 - Model – based algorithm
- Impact assessment of the proposed strategy through simulation
 - Traffic efficiency
 - Safety
 - Environment





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