

Road Safety Analysis using Trajectory Data: At the Intersection of Naturalistic and Simulation Studies

WCTR 2022 virtual event

SIG C4: Traffic Safety Analysis and Policy

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**POLYTECHNIQUE
MONTREAL**

TECHNOLOGICAL
UNIVERSITY



CIRRELT

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Methodology

It Starts with Trajectory Data: Automated Video Analysis

Road User Behaviour and Safety Analysis

Case Studies on Automated Shuttles

Conclusion

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Methods for Road Safety Analysis

There are **two** main categories of methods, whether they are based on **direct observation** or not

1. Accidents are **reconstituted**

- traditional road safety analysis relying on historical collision data
- vehicular accident reconstruction

Methods for Road Safety Analysis

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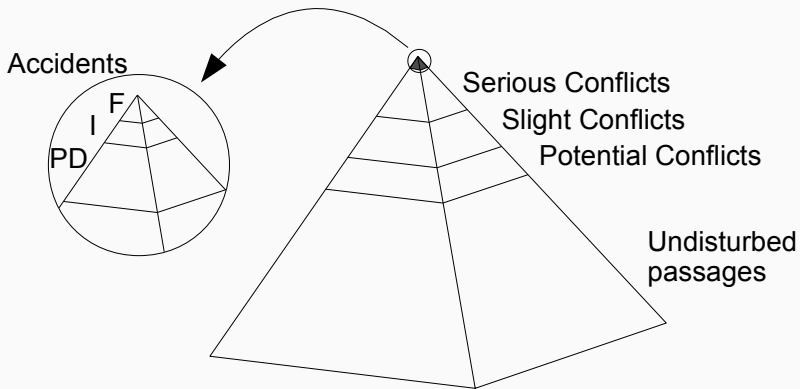
1. Accidents are **reconstituted**

- traditional road safety analysis relying on historical collision data
- vehicular accident reconstruction

2. Road user behavior, interactions and accidents are **directly observed**

- behavioural observations and **surrogate measures of safety (SMoS)**
- data source: naturalistic (driving) studies, probe vehicles, site observations
 - manual to automated collection method

Foundation for Proactive Safety: the Safety Hierarchy



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Processing Steps

1. Video data collection
2. Data preparation
3. Road user detection, tracking and classification

Step 1: Video Data Collection

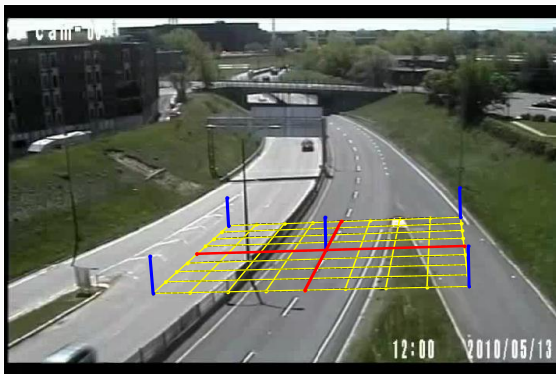
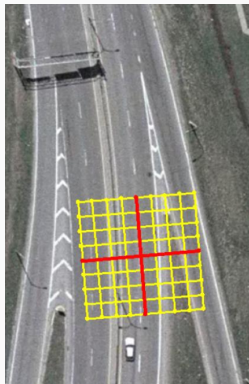


Step 1: Video Data Collection

SAMPLE CAMERA VIEWS UNDER DIFFERENT LIGHTING CONDITIONS					
Daytime Conditions	Thermal Camera	Regular Camera	Nighttime Conditions	Thermal Camera	Regular Camera
Overcast			High visibility		
Sun, no shadows			Medium visibility		
Sun, strong shadows			Low visibility		

Step 2: Data Preparation

In particular, camera calibration: homography, distortion, etc.



Step 2: Data Preparation

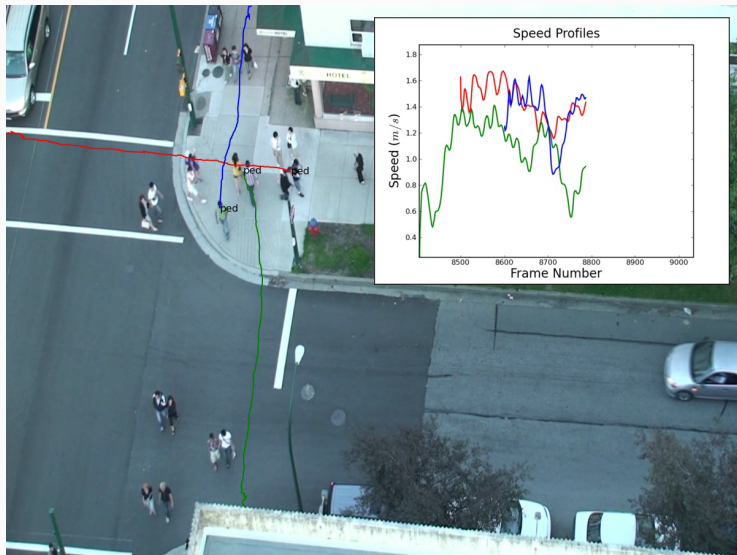
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Step 3: Road User Detection, Tracking and Classification



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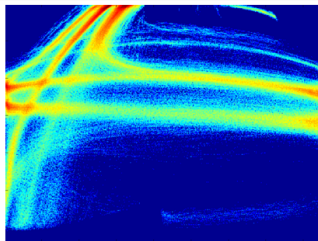


Video

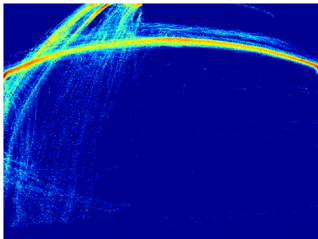
Step 3: Road User Detection, Tracking and Classification



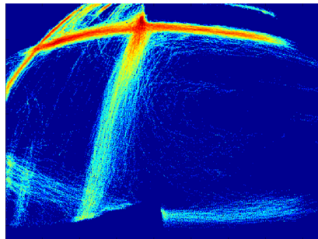
(a) Snapshot of video frame



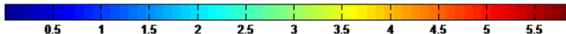
(b) Vehicle trajectory heat-map



(c) Cyclist trajectory heat-map



(d) Pedestrian trajectory heat-map



(e) Scale used for trajectory heat-maps (log-scale)

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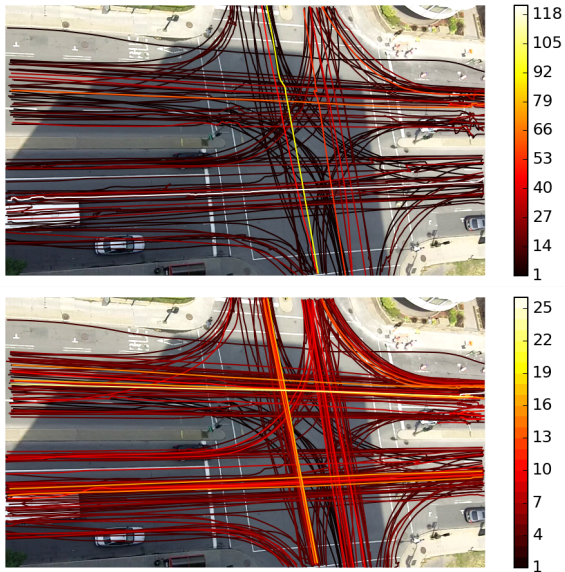
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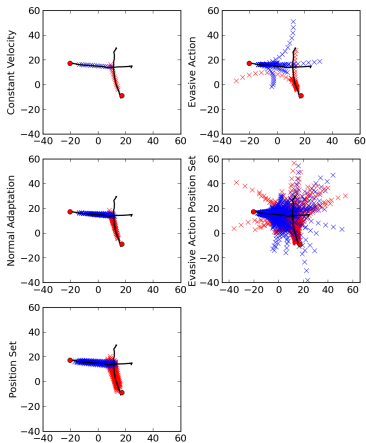
Processing Steps

4. Motion pattern learning
5. Motion prediction
6. Safety indicators
7. Interpretation (SMoS)

Step 4: Motion Pattern Learning

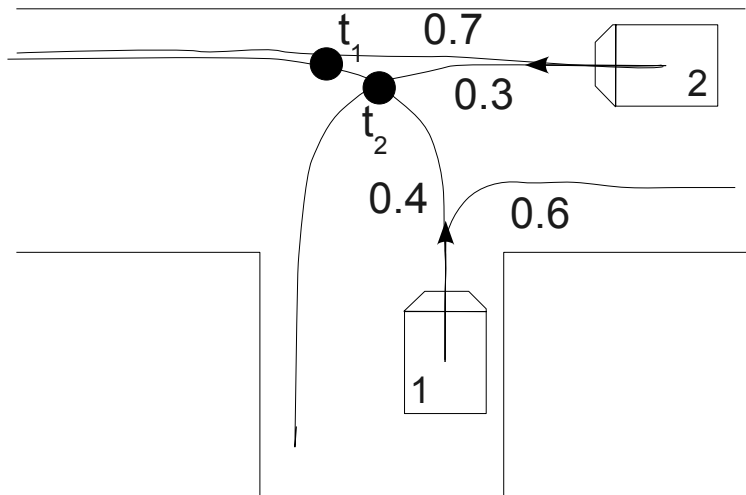


Step 5: Motion Prediction



A traffic conflict is “an observational situation in which two or more road users approach each other in space and time to such an extent that a collision is imminent if their movements remain unchanged”

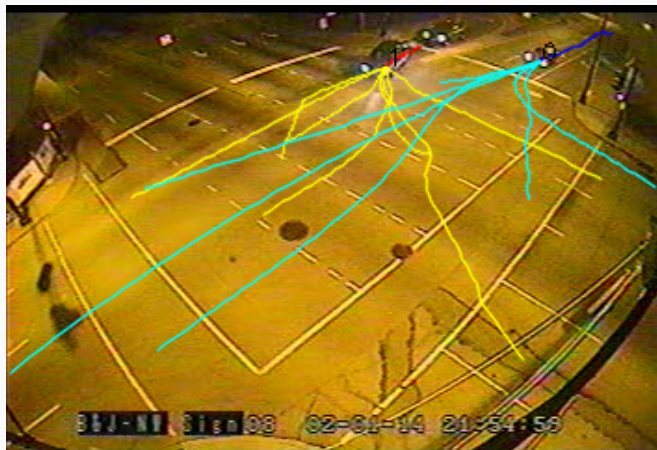
Step 5: Motion Prediction



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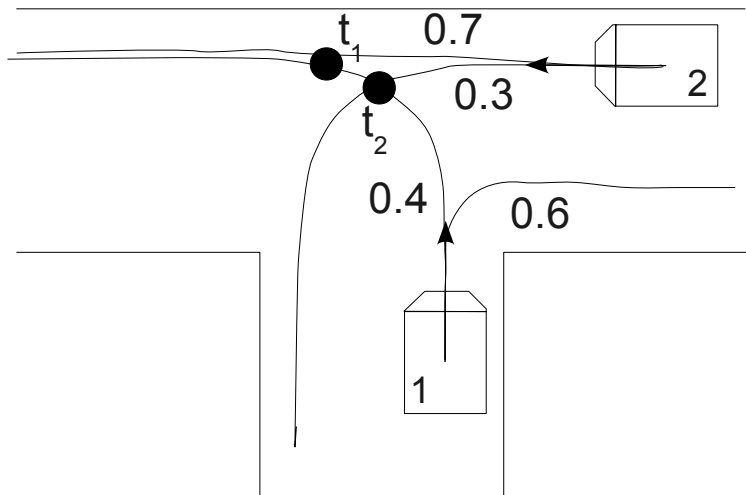
Step 6: Safety Indicators

- **Continuous** measures
 - Time-to-collision (TTC)
 - Gap time (GT) (=predicted PET)
 - Deceleration-based indicators, e.g. deceleration to safety time (DST)
 - Speed-based indicators, (extended) Delta-V , etc.
- **Unique** measure per conflict
 - Post-encroachment time (PET)
 - Evasive action(s) (harshness), subjective judgment, etc.

Step 6: Safety Indicators

- **Continuous** measures (* based on **motion prediction** methods)
 - Time-to-collision (TTC) *
 - Gap time (GT) (=predicted PET) *
 - Deceleration-based indicators, e.g. deceleration to safety time (DST) *
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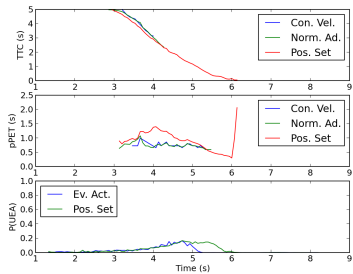
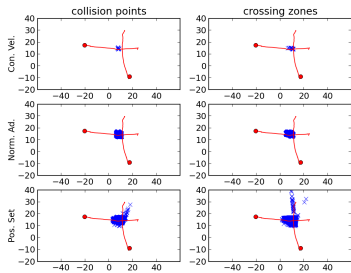
Using of a finite set of predicted trajectories, **enumerate** the collision points CP_n and the crossing zones CZ_m . Safety indicators can then be computed:

$$P(\text{Collision}(U_i, U_j)) = \sum_n P(\text{Collision}(CP_n))$$

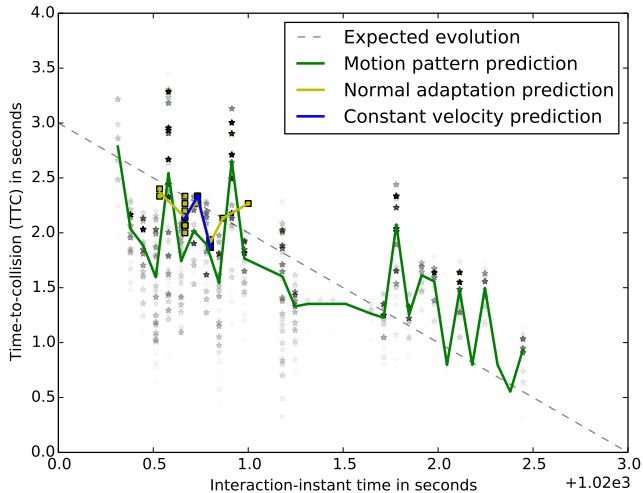
$$TTC(U_i, U_j, t_0) = \frac{\sum_n P(\text{Collision}(CP_n)) t_n}{P(\text{Collision}(U_i, U_j))}$$

$$pPET(U_i, U_j, t_0) = \frac{\sum_m P(\text{Reaching}(CZ_m)) |t_{i,m} - t_{j,m}|}{\sum_m P(\text{Reaching}(CZ_m))}$$

Step 6: Safety Indicators

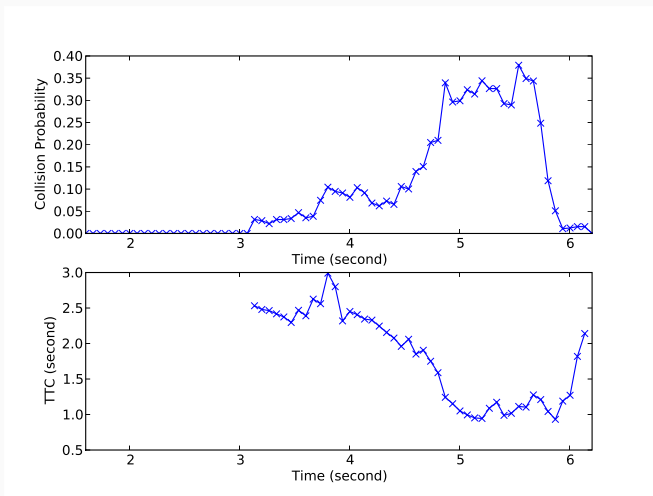


Step 6: Safety Indicators



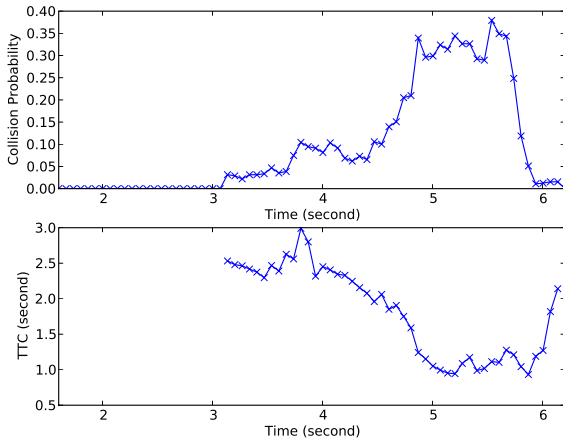
Step 7: Interpretation (SMoS)

For each interaction, we have



Step 7: Interpretation (SMoS)

How should data be aggregated?



Step 7: Interpretation (SMoS)

Should data be **aggregated** (to count severe events)?

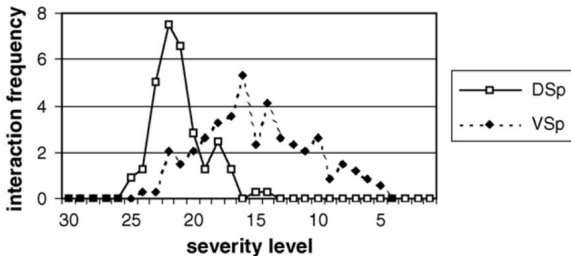
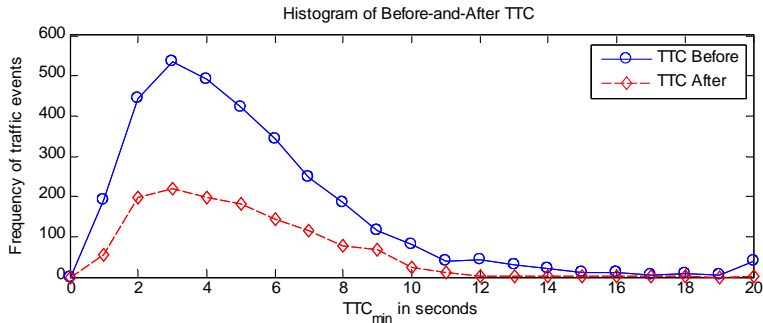


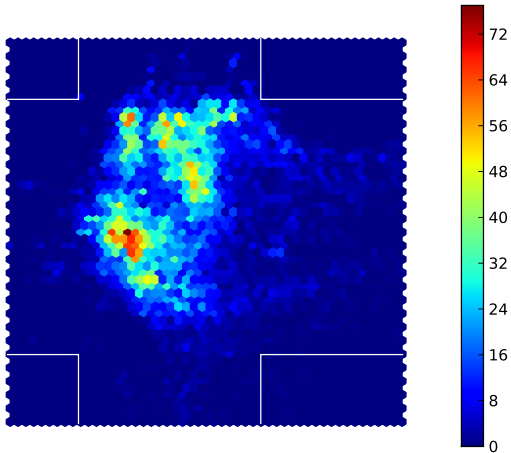
Fig. 6. Interaction frequency (interactions per observation hour) for different severity levels. Straight ahead driving vehicles versus pedestrians. The pedestrian is taking evasive action. A non-signalised intersection (DSp) and a signalised intersection (VSp).

Step 7: Interpretation (SMoS)

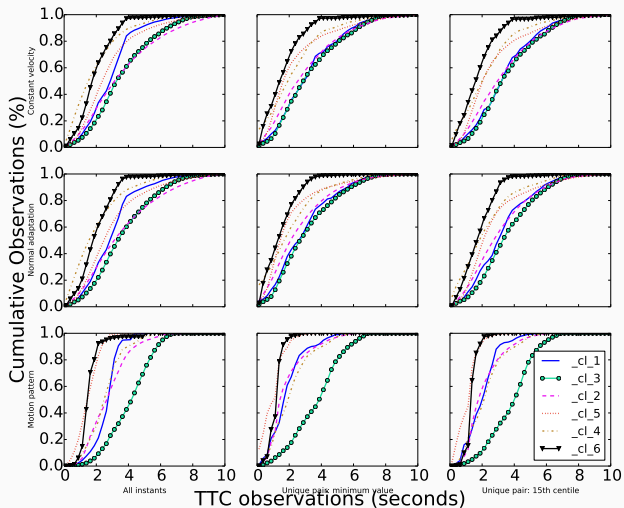


Step 7: Interpretation (SMoS)

Traffic Conflicts



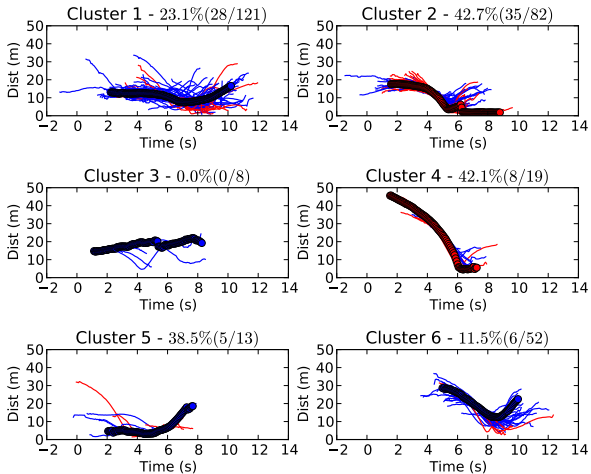
Step 7: Interpretation (SMoS)



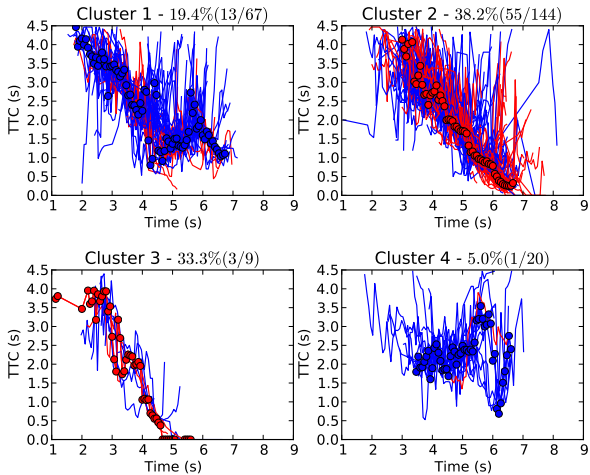
Step 7: Interpretation (SMoS)

	Model I. Cycle track on the right vs. no cycle track			Model II. Cycle track on the left vs. no cycle track			Model III. Cycle track on the right vs. cycle track on the left		
	Coef.	Std. Err.	Sig.	Coef.	Std. Err.	Sig.	Coef.	Std. Err.	Sig.
Cycle Track on Right	0.395	0.181	0.03	-	-	-	-	-	-
Cycle Track on Left	-	-	-	Not Significant			-0.513	0.131	0.00
Bicycle Flow for 5s before to 5s after	Not Significant			0.088	0.038	0.02	0.066	0.034	0.05
Turning-Vehicle Flow for 5s before to 5s after	-2.771	0.132	0.00	-3.265	0.090	0.00	-3.131	0.080	0.00
Number of Lanes on the Main Road	-0.151	0.078	0.05	Not Significant			Not Significant		
Number of Lanes on the Turning Road	Not Significant			0.324	0.146	0.03	0.457	0.178	0.01
Cut-off 1	-6.599	0.353	0.00	-7.372	0.301	0.00	-7.621	0.323	0.00
Cut-off 2	-4.233	0.273	0.00	-3.807	0.223	0.00	-4.125	0.265	0.00
Cut-off 3	-3.150	0.256	0.00	-2.102	0.211	0.00	-2.479	0.258	0.00
Number of Observations	2880			4803			6567		
Log likelihood	-804			-1876			-2330		

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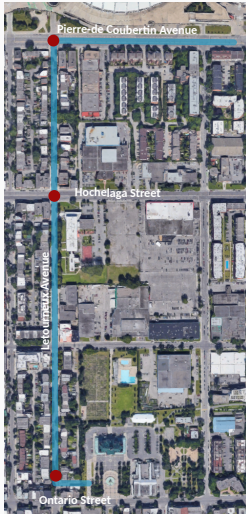
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Safety Studies Using SMOs

- Highway on-ramps and roundabouts
- Cycling infrastructure and the lack thereof
- Pedestrian crosswalks
- Stop-controlled intersections: 2-way vs all-way
- Pedestrian workers (traffic police) vs their stress
- Connected and Automated Vehicles (CAVs)
- Pedestrian countdowns and driver behaviour

Study of Low-Speed Automated Shuttles in Montreal and Candiac



Sites in Montreal

Pierre-de-Coubertin



Hochelaga



Ontario



Sites in Candiac

Residence



Inverness



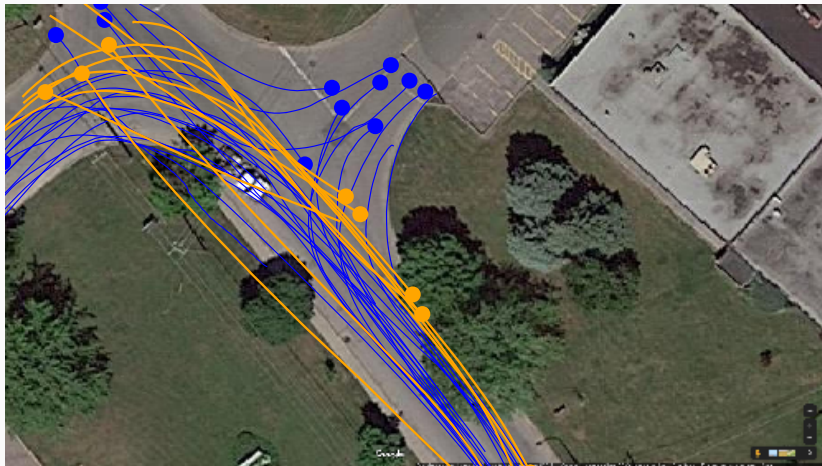
Marie-Victorin



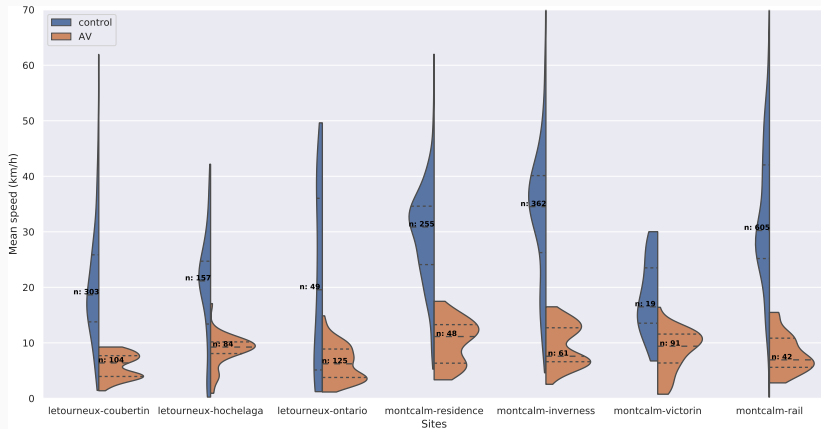
Rail



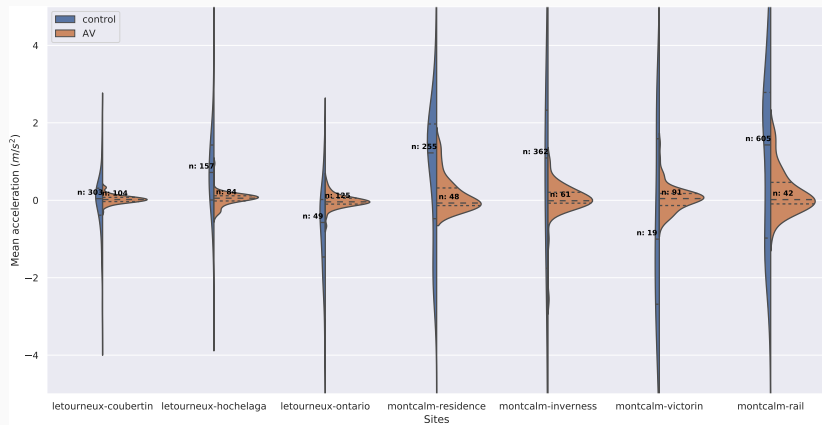
Motion Patterns at Inverness Site



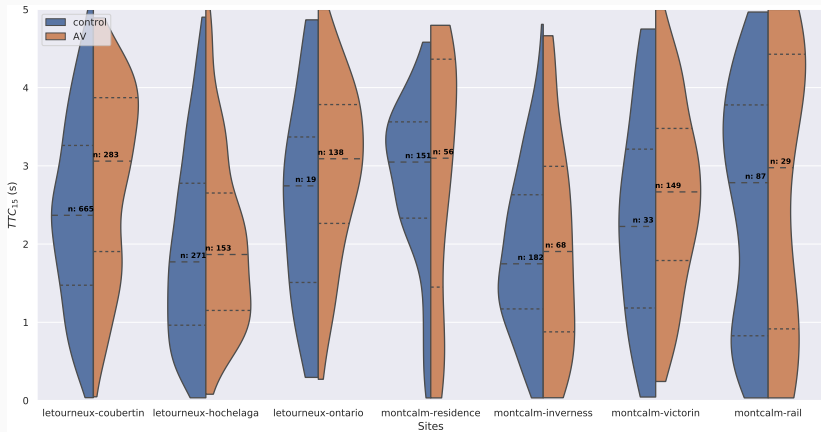
Results



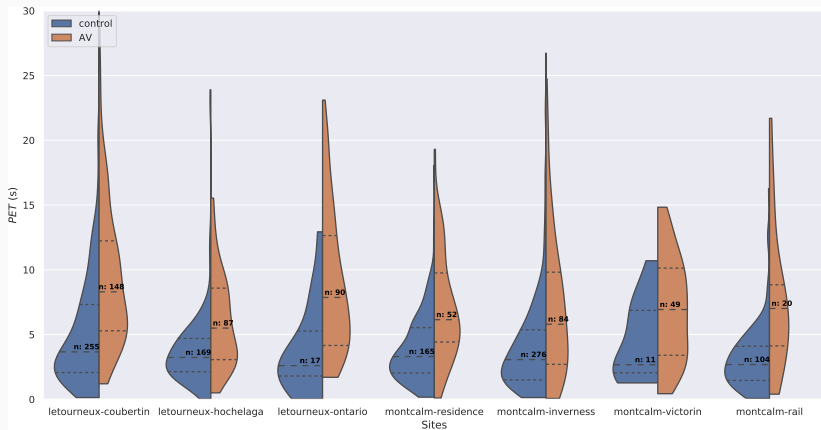
Results



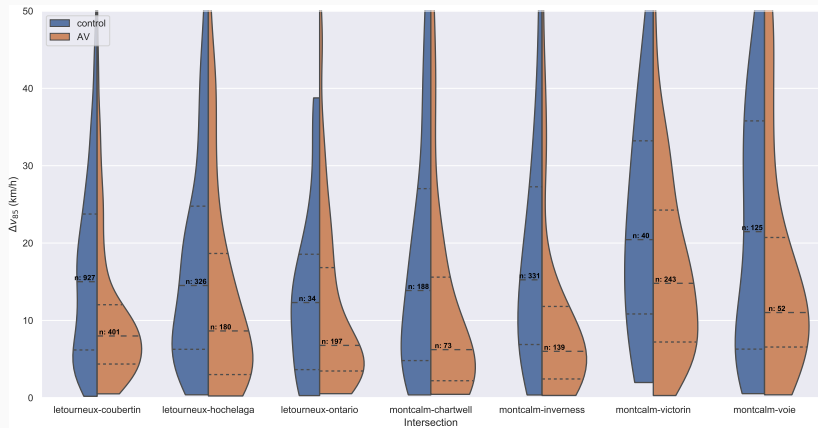
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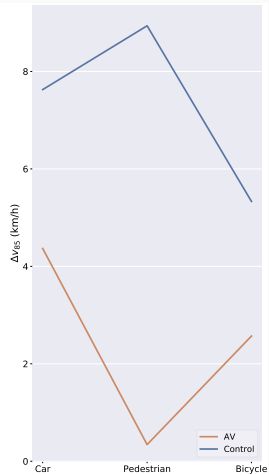
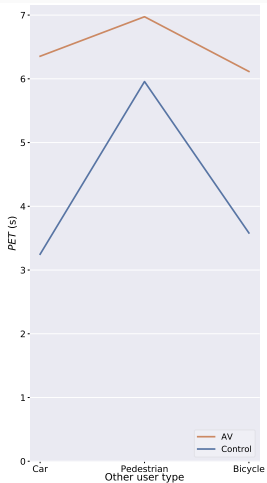
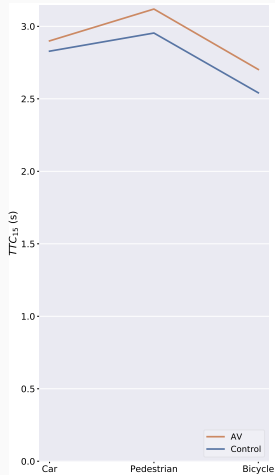
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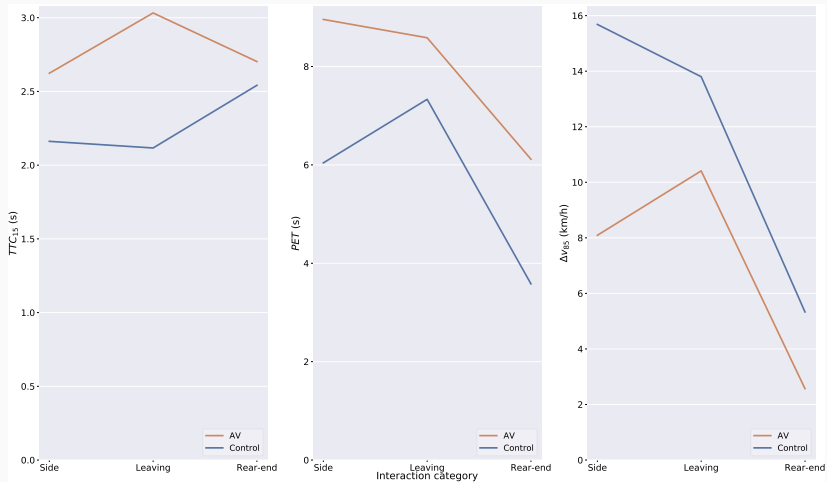
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- **naturalistic** studies: large scale analysis of road user behaviour and safety
- **calibration** of road user models

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We have **access** to increasing amounts of trajectory data, from various sensors (various types of cameras, LIDAR, radar, etc.), which enables

- **naturalistic** studies: large scale analysis of road user behaviour and safety
- **calibration** of road user models
- which **in turn** enable proactive safety analysis, i.e. to predict road safety through the **simulation of potential scenarios at the microscopic level**

Predicting road safety through the simulation of potential scenarios at the microscopic level **to compute safety indicators (e.g. TTC)** is not the same as current traffic simulations that generate SMOs, since, in one way or the other, they **must reproduce the mechanisms that lead to crashes**

Predicting road safety through the simulation of potential scenarios at the microscopic level **to compute safety indicators (e.g. TTC)** is not the same as current traffic simulations that generate SMOs, since, in one way or the other, they **must reproduce the mechanisms that lead to crashes** or you end up **simulating AVs**

It therefore remains an open question

how one should measure safety in a traffic simulation: e.g.
should one simulate conflicts or crashes?

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Questions?