

Automated Proactive Road Safety Analysis

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November 1st 2010

Outline

Motivation

Probabilistic Framework for Automated Road Safety Analysis

Experimental Results using Video Data

Investigating Collision Factors Using Microscopic Data

Conclusion

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A World Health Issue

*Over 1.2 million people die each year on the world's roads, and between 20 and 50 million suffer non-fatal injuries. In most regions of the world this epidemic of road traffic injuries is still **increasing**.*

(Global status report on road safety, World Health Organization, 2009)

A World Health Issue

TOTAL 2004		
RANK	LEADING CAUSE	%
1	Ischaemic heart disease	12.2
2	Cerebrovascular disease	9.7
3	Lower respiratory infections	7.0
4	Chronic obstructive pulmonary disease	5.1
5	Diarrhoeal diseases	3.6
6	HIV/AIDS	3.5
7	Tuberculosis	2.5
8	Trachea, bronchus, lung cancers	2.3
9	Road traffic injuries	2.2
10	Prematurity and low birth weight	2.0
11	Neonatal infections and other	1.9
12	Diabetes mellitus	1.9
13	Malaria	1.7
14	Hypertensive heart disease	1.7
15	Birth asphyxia and birth trauma	1.5
16	Self-inflicted injuries	1.4
17	Stomach cancer	1.4
18	Cirrhosis of the liver	1.3
19	Nephritis and nephrosis	1.3
20	Colon and rectum cancers	1.1

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13	Liver cancer	1.7
14	Colon and rectum cancer	1.7
15	Oesophagus cancer	1.5
16	Violence	1.4
17	Alzheimer and other dementias	1.4
18	Cirrhosis of the liver	1.3
19	Breast cancer	1.3
20	Tuberculosis	1.1

Source: World health statistics 2008 (<http://www.who.int/whosis/whostat/2008/en/index.html>)

Road Safety Analysis

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- Limits of the traditional approach based on historical collision data:
 - Problems of availability and quality
 - Insufficient data to understand the processes that lead to collisions
 - **Reactive** approach
- Need for **proactive** approaches and **surrogate** safety measures that do not depend on the occurrence of collisions

Surrogate Safety Measures

- Research on surrogate safety measures that
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 - are correlated to collisions, logically and statistically
- 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”
[Amundsen and Hydén, 1977]

Automated Methods

- Need for more **microscopic** traffic data
- **Surrogate** safety measures, e.g. in traffic conflict studies, are collected with various levels of automation
- The main bottleneck for proactive methods is that data is still mostly collected **manually**

Outline

Motivation

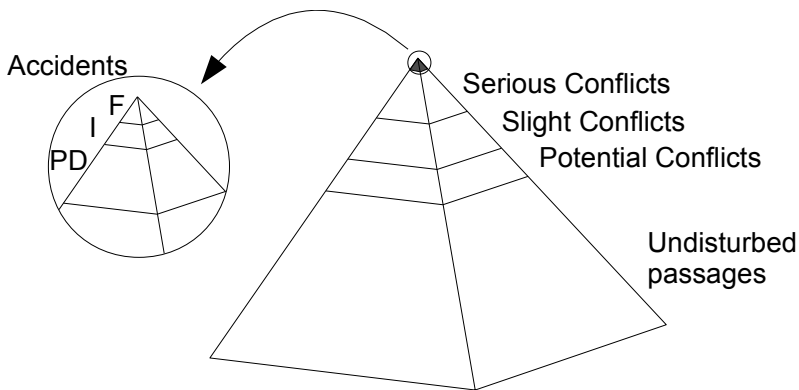
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The Safety/Severity Hierarchy



Various severity measures

The Collision Course

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The Collision Course

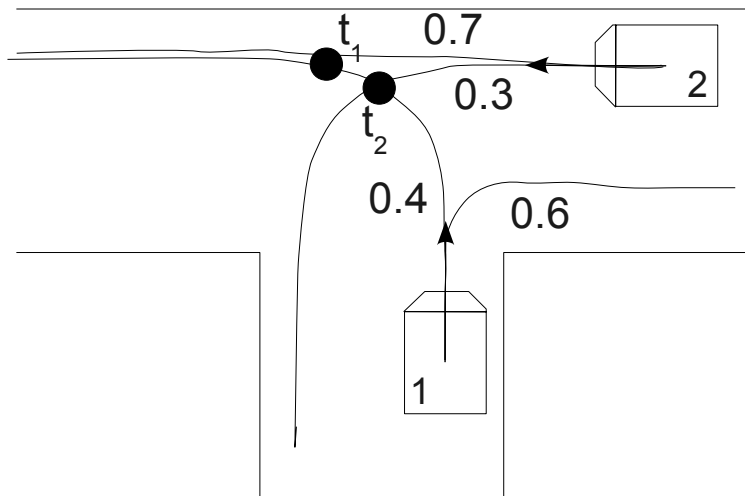
- A traffic conflict is “an observational situation in which two or more road users approach each other in space and time to such an extent that a collision is imminent **if their movements remain unchanged**”
- For two interacting road users, **many** chains of events may lead to a collision
- It is possible to estimate the probability of collision if one can **predict** the road users' future positions

Movement Prediction

- Learn road users' **motion patterns** (including frequencies), represented by actual trajectories called **prototypes**
- Match observed trajectories to prototypes and extrapolate

[Saunier et al., 2007, Saunier and Sayed, 2008]

A Simple Example



Collision Points

Using of a finite set of extrapolation hypotheses, **enumerate** the collision points CP_n . Severity indicators can then be computed:

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

[Saunier et al., 2010]

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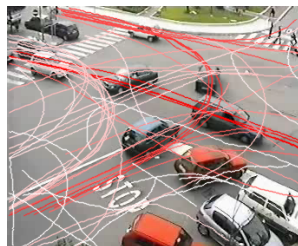
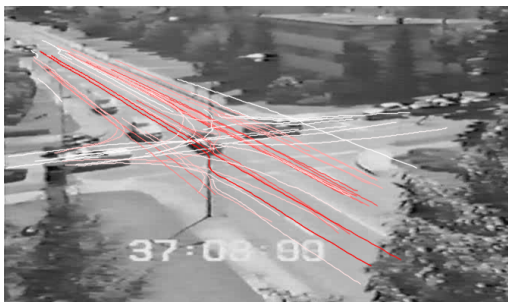
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Motion Pattern Learning

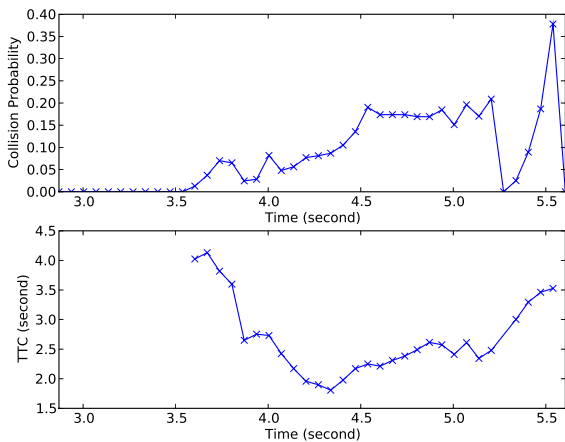


Traffic Conflict Dataset, Vancouver	Reggio Calabria, Italy
58 prototype trajectories (2941 trajectories)	58 prototype trajectories (138009 trajectoires)

The Kentucky Dataset

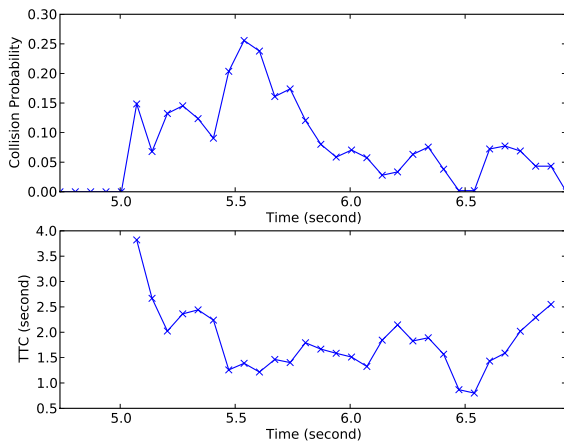
- Video recordings kept for a few seconds before and after the sound-based automatic detection of an interaction of interest
 - 229 traffic conflicts
 - 101 collisions
 - The existence of an interaction or its severity is not always obvious
 - The interactions recorded in this dataset involve only motorized vehicles
 - Limited quality of the video data: resolution, compression, weather and lighting conditions
- Calibration done using the tool developed by Karim Ismail at UBC [Ismail et al., 2010b]

Severity Indicators



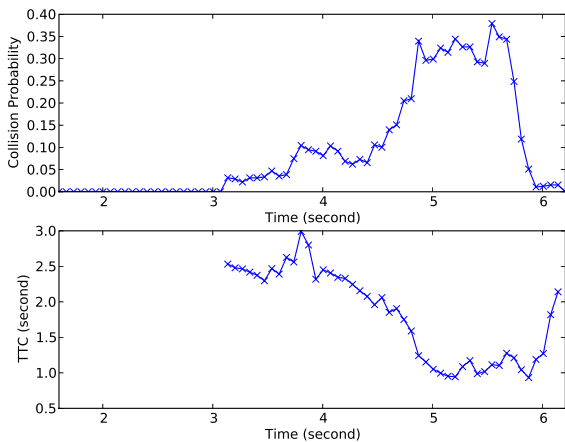
Side conflict

Severity Indicators



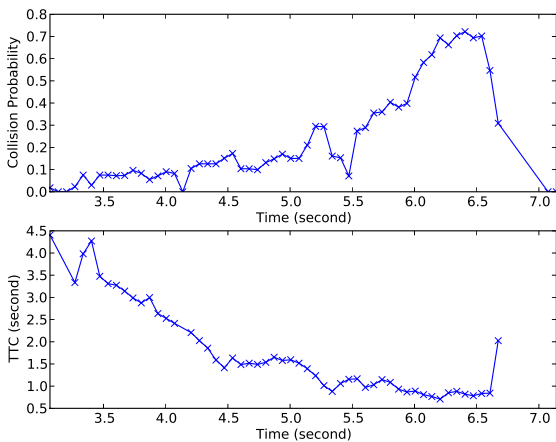
Side conflict

Severity Indicators



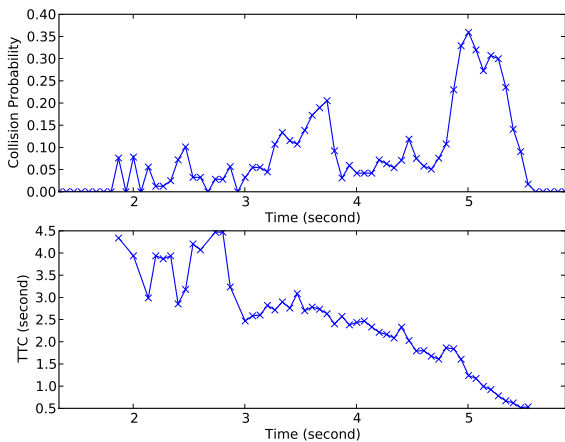
Parallel conflict

Severity Indicators



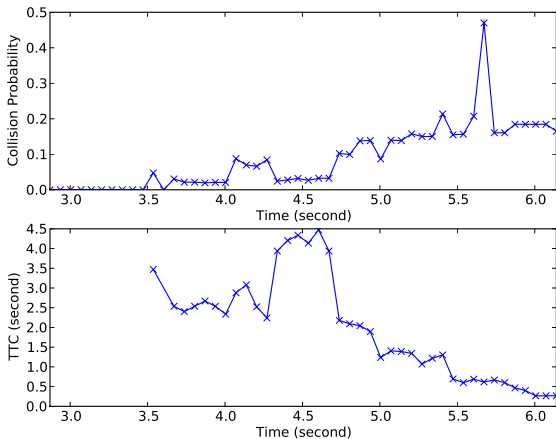
Side collision

Severity Indicators



Side collision

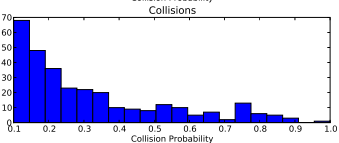
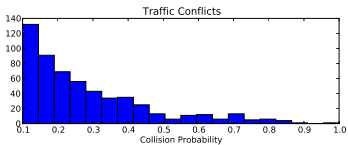
Severity Indicators



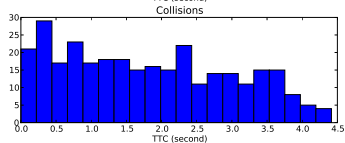
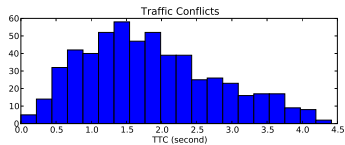
Parallel collision

Distribution of Indicators

Maximum Collision Probability

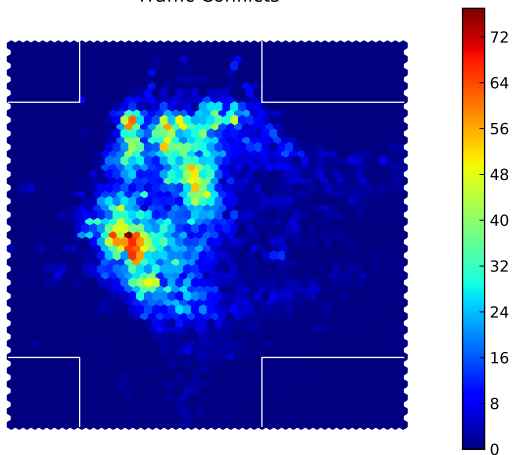


Minimum TTC

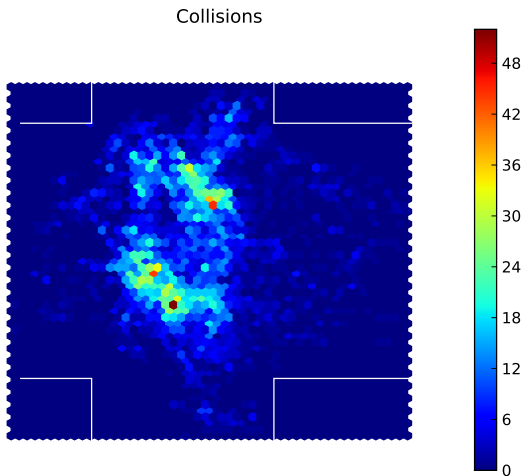


Spatial Distribution of the Collision Points

Traffic Conflicts



Spatial Distribution of the Collision Points

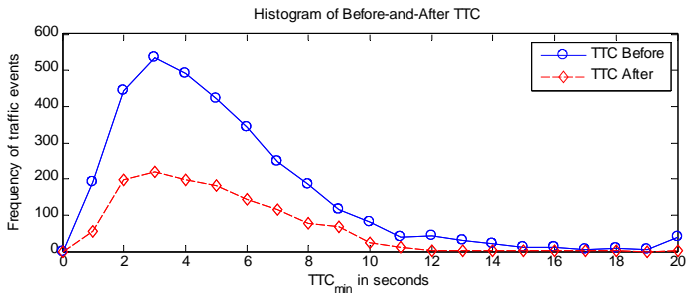


Study Before and After the Introduction of a Scramble Phase

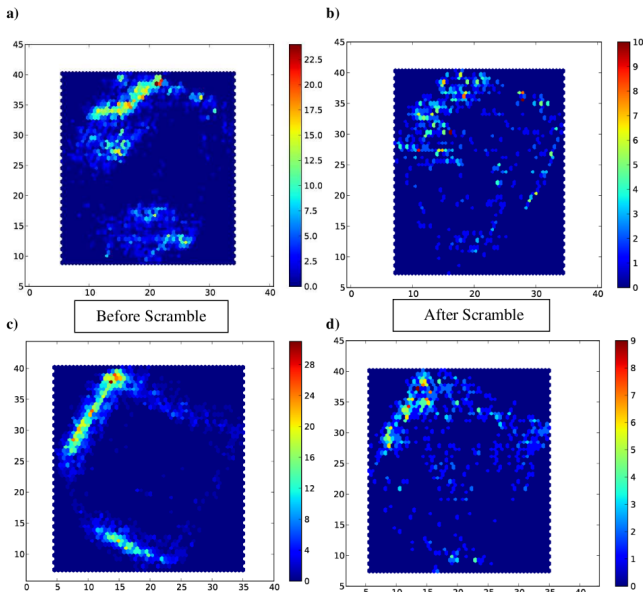


Data collected in Oakland, CA [Ismail et al., 2010a]

Distribution of Severity Indicators



Before and After Distribution of the Collision Points



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- Understand **collision processes** by studying the similarities of interactions with and without a collision (conflicts)

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 - Importance for surrogate safety measures: what interactions without a collision may be used as surrogates for collisions?

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 - Importance for surrogate safety measures: what interactions without a collision may be used as surrogates for collisions?
- Use of **data mining** techniques (k-means and hierarchical agglomerative clustering method) to cluster the data

[Saunier et al., 2011]

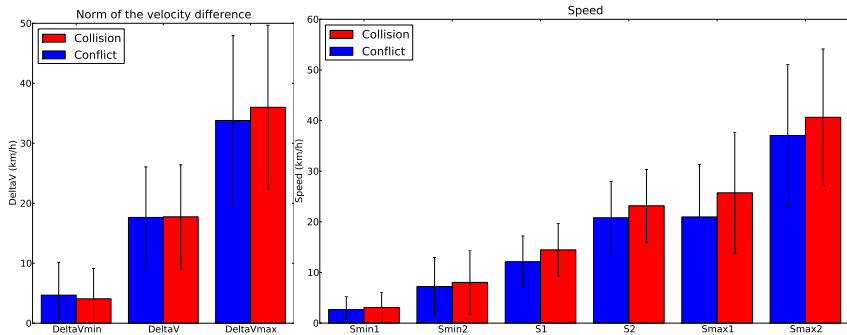
Description of Interactions

Categorical attributes	Values
<i>Type of day</i>	weekday, week end
<i>Lighting condition</i>	daytime, twilight, nighttime
<i>Weather condition</i>	normal, rain, snow
<i>Interaction category</i>	same direction, opposite direction, side
<i>Interaction outcome</i>	conflict, collision

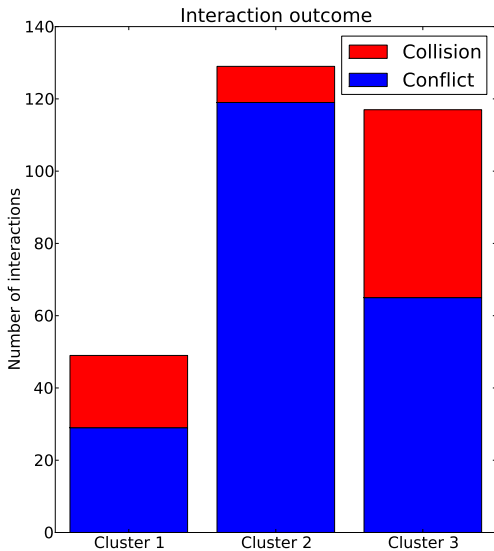
Description of Interactions (continued)

Numerical attributes	Units
<i>Road user type</i>	
passenger car	number of road users
van, 4x4, SUV	number of road users
bus	number of road users
truck (all sizes)	number of road users
motorcycle	number of road users
bike	number of road users
pedestrian	number of road users
<i>Type of evasive action</i>	
No evasive action	number of evasive actions
Braking	number of evasive actions
Swerving	number of evasive actions
Acceleration	number of evasive actions
<i>3 attributes from Δv</i>	km/h
<i>6 values from s</i>	km/h

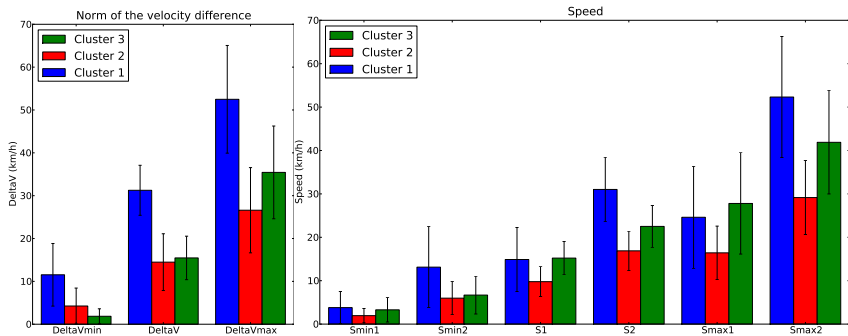
Distribution of Speed Attributes



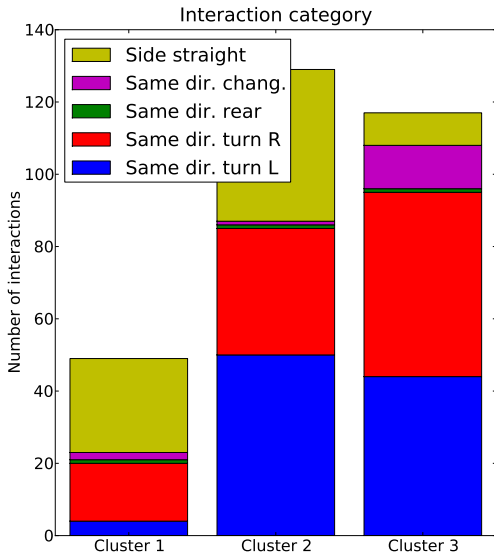
3 Clusters



3 Clusters: Speed Attributes



3 Clusters: Interaction Category



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- Future work:
 - Still more work on data collection techniques (computer vision)
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 - Understanding and modelling of collision processes: collect more data
- Need for more **open** science: data and code sharing
<http://nicolas.saunier.confins.net>

Collaboration with

- Clark Lim and Tarek Sayed (University of British Columbia)
- Karim Ismail (Carleton University)
- Nadia Mourji, Bruno Agard (École Polytechnique de Montréal)

Questions ?

-  Amundsen, F. and Hydén, C., editors (1977).
Proceedings of the first workshop on traffic conflicts, Oslo, Norway. Institute of Transport Economics.
-  Davis, G. A., Hourdos, J., and Xiong, H. (2008).
Outline of causal theory of traffic conflicts and collisions.
In Transportation Research Board Annual Meeting Compendium of Papers.
08-2431.
-  Ismail, K., Sayed, T., and Saunier, N. (2010a).
Automated analysis of pedestrian-vehicle conflicts: A context for before-and-after studies.
Transportation Research Record: Journal of the Transportation Research Board.
In press.
-  Ismail, K., Sayed, T., and Saunier, N. (2010b).
Camera calibration for urban traffic scenes: Practical issues and a robust approach.

*In Transportation Research Board Annual Meeting
Compendium of Papers, Washington, D.C.
10-2715.*



Saunier, N., Mourji, N., and Agard, B. (2011).

Investigating collision factors by mining microscopic data of vehicle conflicts and collisions.

*In Transportation Research Board Annual Meeting
Compendium of Papers, Washington, D.C.
11-0117.*



Saunier, N. and Sayed, T. (2008).

A Probabilistic Framework for Automated Analysis of Exposure to Road Collisions.

*Transportation Research Record: Journal of the
Transportation Research Board, 2083:96–104.*



Saunier, N., Sayed, T., and Ismail, K. (2010).

Large scale automated analysis of vehicle interactions and collisions.

Transportation Research Record: Journal of the Transportation Research Board, 2147:42–50.



Saunier, N., Sayed, T., and Lim, C. (2007).

Probabilistic Collision Prediction for Vision-Based Automated Road Safety Analysis.

In The 10th International IEEE Conference on Intelligent Transportation Systems, pages 872–878, Seattle. IEEE.