



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

Spatial analysis of telematics surrogate safety measures across road environments

Dimitrios Nikolaou

Transportation Engineer, PhD Candidate

Together with: Armira Kontaxi, Apostolos Ziakopoulos, George Yannis



National Technical University of Athens

The SmartMaps project

- Project partners:
 - National Technical University of Athens, Department of Transportation Planning and Engineering <u>www.nrso.ntua.gr</u>
 - OSeven Telematics <u>www.oseven.io</u>
 - Global Link <u>www.globallink.gr</u>
- > Duration of the project:
 - 30 months (June 2021 November 2023)
- Operational Program:
 - "Competitiveness, Entrepreneurship and Innovation" (EPAnEK) of the National Strategic Reference Framework (NSRF) – 2nd iteration









ΕΛΛΗΝΙΚΗ ΔΗΜΟΚΡΑΤΙΑ ΥΠΟΥΡΓΕΙΟ ΠΑΙΔΕΙΑΣ ΈΥΝΑΣ & ΟΡΗΣΚΕΥΜΑΤΩΝ ΕΛΛΗΝΙΚΗ ΔΗΜΟΚΡΑΤΙΑ ΥΠΟΥΡΓΕΙΟ ΟΙΚΟΝΟΜΙΑΣ & ΑΝΑΠΤΥΞΗΣ ΕΙΔΙΚΗ ΓΡΑΜΜΑΤΕΙΑ ΕΠΑ & ΤΣ ΕΙΔΙΚΗ ΥΠΗΡΕΣΙΑ ΔΙΑΧΕΙΡΙΣΗΣ ΕΠΑΥΕΚ

ENESS• ENTREPRENEURSHIP•INNOVATION

Objectives

Exploitation of large-scale spatio-temporal data from smartphone sensors.

- Development of smart driver behaviour maps with online information on safety conditions and eco-driving (by reducing fuel consumption).
- Creation of a comprehensive tool to promote safe driving behaviour with application in Greece and around the world.





Data Collection

Road Geometry Data (OpenStreetMap)

- ➢ Length
- > Curvature
- > Slope

Observed Driving Data – Field (Global Link)

- Seatbelt use
- > Helmet use
- Speeding
- Distraction

Naturalistic Driving Data – Telematics (OSeven)

- ➢ Harsh braking
- Harsh acceleration
- ➤ Speeding
- Distraction

Road Crash Data (ELSTAT)





Road Geometry Data

- The area of East Macedonia & Thrace was chosen as a challenging area in terms of data availability for the initial investigations.
- The process of data collection and analysis carried out in East Macedonia & Thrace area will be replicated in the remaining Greek Regions.
- 6099 road segments: (Mean Length: 290m, Mean Angle Rate: 0.50 [1/m], Total Length 1700km)
- Road Types: (68% residential, 12% tertiary, 7% secondary, 3% motorway, 10% other types)
- Slopes: 76% (flat: 0-3%), 10% (mild: 3-5%), 7% (medium: 5-8%), 3% (hard: 8-10%), 4% (extreme: >10%).





Observed Driving Data

- Field measurements on road user behaviour indicators in 10 locations (3 motorway, 4 rural, 3 urban).
- Inverse Distance Weighting (IDW) was used twice for spatial interpolation in the entire road network (motorways, non-motorways).
- IDW estimates the value of a variable at a given location by using a weighted average of the surrounding known values, with weights determined by their distance to the target location, assuming that nearby locations have similar values.
- ~3500 observations of passenger car drivers. (seatbelt, distraction, speeding)
 - ~260 observations of PTW drivers (helmet).



Naturalistic Driving Data - Telematics

> 5129 trips in the examined area in 2021.

Map matching of naturalistic driving data and considered road segments.

Naturalistic Driving Data per segment	Min.	Mean	Max.
Trip count	0	32	1272
Speeding rate (sec/trips)	0	0.26	110
Mobile usage rate (sec/trips)	0	0.34	133
Harsh acceleration rate (sec/trips)	0	0.004	1.00
Harsh braking rate (sec/trips)	0	0.007	1.42





Road Crash Data

- Inaccurate recording of crash locations (lack of coding with geographical coordinates in the national database).
- Aggregate crash data for 5 municipalities (Xanthi, Avdira, Myki, Topeiros, Nestos)

	20)
Xanthi No 108 10 15 128 25	
Xanthi Yes 1 0 2 0 2	
Avdira No 70 12 20 73 32	
Avdira Yes 5 1 2 10 3	
Myki No 20 4 6 16 10	
Myki Yes 0 0 0 0 0	
Topeiros No 32 7 7 36 14	
TopeirosYes43477	
Nestos No 44 12 9 49 21	
Nestos Yes 12 3 4 15 7	

Spatial interpolation of "area crash-related indexes" based on the total numbers using IDW twice (motorways, non-motorways). (Crashes, Fatalities, Killed and Seriously Injured (KSI)).





Spatial Error Model - Background

- The spatial error model handles the spatial autocorrelation in the residuals.
- The idea is that such errors (residuals from regression) are autocorrelated in that the error from one spatial feature can be modeled as a weighted average of the errors of its neighbors.
- > This model can be expressed as:
 - $y = X\beta + u$, $u = \lambda_{Err} Wu + \epsilon$
 - where y is an (N×1) vector of observations on a response variable taken at each of N locations,
 - > X is an (N×k) matrix of covariates,
 - \succ β is a (k×1) vector of parameters,
 - \succ u is an (N×1) spatially autocorrelated disturbance vector,
 - ε is an (N×1) vector of independent and identically distributed disturbances
 - \succ λ_{Err} is a scalar spatial parameter.



Spatial Error Model - Results

Dependent variable: log(harsh_braking_count + 1)

Type: error

Coefficients: (asymptotic standard errors)

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.3077	0.0849	-3.6229	0.0003
log(1 + length)	0.0360	0.0039	9.3512	< 2.2e-16
log(1 + slope)	-0.0055	0.0059	-0.9399	0.3473
log(1 + efficiency)	0.1067	0.0580	1.8408	0.0657
log(1 + speeding_count)	0.0847	0.0046	18.4841	< 2.2e-16
mobile_usage_rate	0.0064	0.0015	4.3898	<0.001
PC_D_Seatbelt_Yes_p	-0.0237	0.0951	-0.2493	0.8031
log(Crashes2016_2020)	0.0233	0.0085	2.7405	0.0061
trip_count	0.0024	0.0000	52.0655	< 2.2e-16

Lambda: 0.022785, LR test value: 5.1845, p-value: 0.022789 AIC: 3891.5, (AIC for Im: 3894.7)

> Lambda value of 0.022 is statistically significant, suggesting the error term is spatially autoregressive.

From the AIC, the spatial error model performs much better than the linear model, as lower AIC indicates better fit.



Spatial Lag Model - Background

- A spatial lag is a variable that essentially averages the neighboring values of a location (the value of each neighboring location is multiplied by the spatial weight and then the products are summed).
- It can be used to compare the neighboring values with those of the location itself.
- For these spatial lags, we can use the spatial lag model to address the spatial autocorrelation in the dependent variable:

 $y = \rho_{Lag}Wy + X\beta + \epsilon$

Where ρ_{Lag} is a scalar spatial parameter, indicating how much a spatial feature is influenced by its neighbors..







Spatial Lag Model - Results

Dependent variable: log(harsh_braking_count + 1)

Type: lag

Coefficients: (asymptotic standard errors)

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.2929	0.0845	-3.4657	0.0005
log(1 + length)	0.0356	0.0038	9.2738	< 2.2e-16
log(1 + slope)	-0.0061	0.0059	-1.0367	0.2999
log(1 + efficiency)	0.1048	0.0579	1.8093	0.0704
log(1 + speeding_count)	0.0846	0.0046	18.4784	< 2.2e-16
mobile_usage_rate	0.0064	0.0015	4.4202	<0.001
PC_D_Seatbelt_Yes_p	-0.0371	0.0944	-0.3927	0.6945
log(Crashes2016_2020)	0.0218	0.0085	2.5734	0.0101
trip_count	0.0024	0.0000	51.9864	< 2.2e-16

Rho: 0.016061, LR test value: 3.8597, p-value: 0.04946 AIC: 3892.8, (AIC for Im: 3894.7)

Rho value of 0.016 is statistically significant, suggesting there is positive spatial autocorrelation.

Similarly, the spatial lag model is much better than the linear model, even though it is not as good as the spatial error model.



Key Conclusions

- Road geometry characteristics, naturalistic driving data, observed driving data and historical road crashes were combined for road safety modelling.
- Significant positive effects of segment length, speeding events, and trip count on harsh braking events count.
- Spatial models provide a better fit to the data than non-spatial models.
- Methodology applied in East Macedonia & Thrace area can be extended to other Greek regions and national road network.









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

Spatial analysis of telematics surrogate safety measures across road environments

Dimitrios Nikolaou

Transportation Engineer, PhD Candidate

Together with: Armira Kontaxi, Apostolos Ziakopoulos, George Yannis



National Technical University of Athens