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**ABSTRACT**

Injuries are one of the main causes of death according to W.H.O. For this reason the attention of road safety researchers especially regards the study of the relationship between driver and road environment.

Several research works show that operating speed is an excellent driver behavior parameter. This article describes a different approach to the classical definition of prediction models for operating speed on horizontal curves. In this paper, the fundamental theories, the applied operating procedures and the first results obtained with the application of Geostatistics are discussed. The mathematical models expressing operating speed in function of horizontal curves characteristics found in International scientific literature, have mainly been built on the basis of Classical Statistics. For this reason, it needs to be pointed out that the interpolative techniques found in Classical Statistics are based upon the use of canonical forms (linear or polynomial regressions) that completely ignore the correlation law between collected data. As such, the determined interpolation stems from the assumption that the data represent a random sample.

The models described in this article have instead been created with the geostatistical interpolation technique (i.e. Kriging). This technique allows to obtain the "*best*" estimates possible because it considers the true correlation law between the measured data.

The applied methods are then described along with the results obtained in the field of road safety by applying Geostatistics which, for several years, have been used, with positive results, in all scientific and engineering fields dealing with empirical data analysis and processing.

## 1 INTRODUCTION

2  
3 Numerous studies have been developed in the last decades in the field of road safety in order  
4 to forecast, already during the project phase, the relationship between real driver behavior and rural  
5 road features (especially on horizontal curve, which are more dangerous than tangent). Many of these  
6 studies focus on creating analytical models expressing driver behavior in relation to the characteristics  
7 of the two lane rural roads. In these models, driver behavior is expressed by the operating speed ( $V_{85}$ ,  
8 that represents the 85th-percentile speed), while the road characteristics are generally represented by  
9 the Curvature Change Rate of the single curve (CCRs) or the horizontal curve radius ( $R$ ) (1).

10 To create such models, various research groups have carried out a lot of survey campaigns to  
11 collect a numerous mass of inherent data on the characteristics of the road and the corresponding  
12 operating speed, to elaborate with statistical techniques (2).

13 The present article arises from these considerations with the objective of introducing the  
14 experiences accomplished by our research team in the developing new analytical models to predict  
15  $V_{85}$ . These new predict models have been created by Geostatistics analysis and interpolation. This  
16 technique among the various advantages offered, allows to obtain the interpolation that minimizes the  
17 estimated variance value: to determine unknown values, it considers only the experimental values that  
18 exercise a real influence and weighs them with appropriate coefficients obtained by studying the  
19 correlation laws between experimental values themselves. The strong point of this approach is to  
20 admit that the collected data do not represent a sample of random values but they are a punctual  
21 manifestation of a phenomenological reality regulated by a precise law. This is a strong point of the  
22 Geostatistics approach in respect to the Statistics one.

23 The article is divided into four main parts. The first part briefly summarizes the main fields of  
24 application of Geostatistics and the advantages of its use, reporting a schematic description of the  
25 fundamental steps making up the geostatistical analysis. The second part describes the data collected  
26 to calibrate the proposed model. The third part describes the development of our proposed predict  
27 model using geostatistical techniques. The final part summarizes the main conclusions of the study  
28 and briefly indicates its future developments.

## 30 GEOSTATISTICAL ANALYSIS OF EXPERIMENTAL DATA

31  
32 In all scientific fields, experimental data analysis and processing assume a considerable  
33 importance. In fact, in the majority of cases, starting from a limited set of collected information it is  
34 necessary to reconstruct a phenomenon of interest within its existing domain. In order to do that, the  
35 enlargement of the sampling mesh does not represent the best solution: investigation time and costs or,  
36 in some cases, the inaccessibility of points of measurement can hinder the feasibility of this choice.  
37 Therefore, the target is, in practice, to maintain low (as possible) the number of observations and to  
38 proceed to create representative models of the phenomenon in study.

39 Among the available alternatives, Geostatistics analytical and interpolation techniques have  
40 assumed a high importance today in all those branches of engineering science requiring experimental  
41 data process instruments. These techniques, now implemented in all most common software (for  
42 example, Surfer and ArcGIS), are based on correlation studies of collected observations and use of  
43 that law for the interpolation.

44 This is because the experimental data gathered may not represent a random sample of values  
45 but, in interpreting the precise expression of the same phenomenon, must be objectively related to  
46 each other. Recognition of the correlation law between the observations allows to carry out an  
47 interpolation based on the real law of phenomenon.

48 Geostatistics was born last century in the mining field, based on studies conducted by Georges  
49 Matheron, Danie Krige and Herbert Sichel, as science for providing tools for the evaluation of mineral  
50 deposits. Since then a lot of progresses has been made in the development of Geostatistics techniques

1 of analysis and interpolation. For this reason Geostatistics has become an extremely powerful tool for  
 2 studying and modeling of space/time phenomena.

3 Therefore Geostatistics supplies a collection of techniques addressed to the study of spatial  
 4 correlation between experimental values of a specific variable, representing a phenomenon in study in  
 5 order to determine the value of the unknown points within the existing dominion (3) (4) (5).

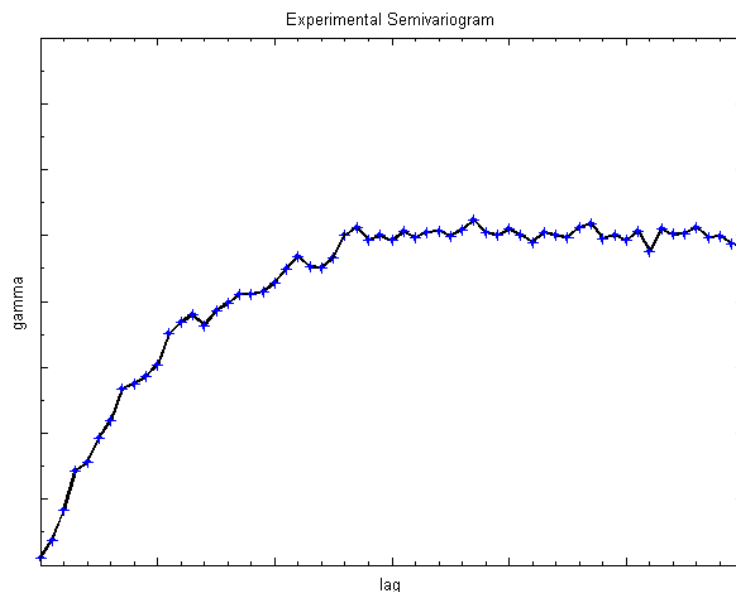
6 In the research work proposed in this paper, the classical Geostatistics tools are used in an  
 7 innovative way for studying the correlation law between the operating speed and road parameters (as  
 8 CCRs or R) that are not a space and/or time coordinates but only design features.

9 More simply, a geostatistical analysis consists of the following main steps:

10 **1. estimate of spatial correlation between measured observations.** Such analysis is carried  
 11 out by construction of an experimental curve (Figure 1) expressing the spatial and/or time correlation  
 12 between collected data. This relationship, or more precisely, the variability of the size studied is  
 13 represented by the function of the experimental semivariogram (Figure 1), defined by the equation:

$$14 \quad 15 \quad 16 \quad 17 \quad 18 \quad \gamma(h) = \frac{1}{2} \text{Var}[Z(x+h) - Z(x)]^2 \quad (1)$$

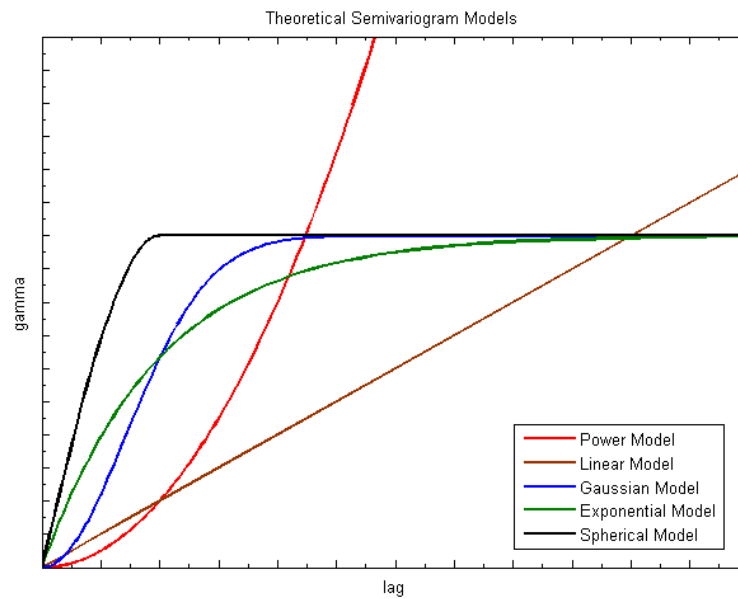
19  $\gamma(h)$  represents the semi-variance between collected pairs of observations separated by a specific  
 20 distance (measured along the horizontal axis) or, rather, the error which would be made if the  
 21 unknown point was replaced with the collected data at the definite distance (4).



**FIGURE 1 An experimental semivariogram example**

42 **2. modeling the experimental semivariogram using a theoretic model,** of a well-known  
 43 equation, that fits the trend of experimental semivariogram which, as already stated, represents the  
 44 phenomenon law. Calculating and modeling the experimental semivariogram offers a series of  
 45 important information of the phenomenon in examination, but it also represents the need of a basis to  
 46 define the values of the unknown points.

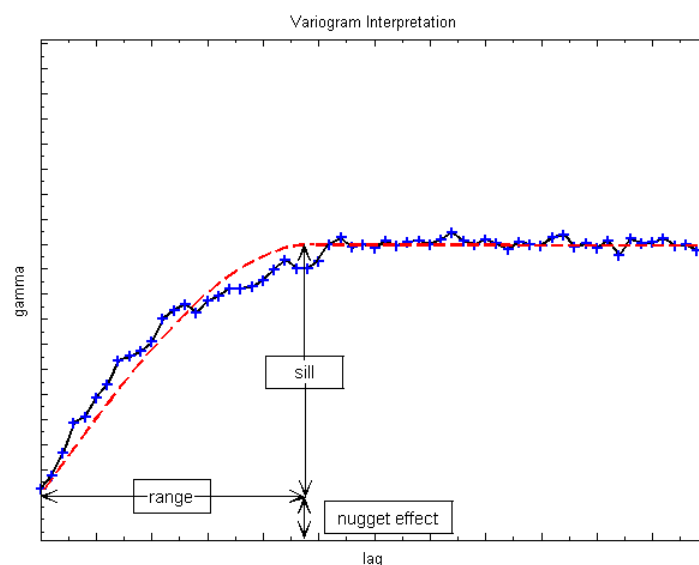
47 In the Figures 2 some of the main theoretical models are shown.



20 **FIGURE 2 Theoretical models for experimental semivariogram interpretation**

21  
22 3. **definition of the variable parameters of the phenomenon in study.** This step is  
23 necessary to identify uniquely the phenomenon law in order to obtain the best fit between the  
24 experimental and theoretical semivariogram. Referring to the Figure 3, these parameters are:

- 25
- 26 • the **nugget effect**, which represents the value of the semivariogram for pairs of observations separated from a distances close to zero;
  - 27 • the **range**, which represents the distance, measured along the horizontal axis, where the curve of the semivariogram tends to flatten, or rather, beyond which the data become totally independent: for this reason all values beyond this limit in respect to the unknown point, should not be considered during the interpolation;
  - 28 • the **sill** corresponds to the value of the semivariogram which is reached at the distance equal to the range.
- 29  
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**FIGURE 3 Variographic parameters of experimental semivariogram**

1  
2  
3 **4. estimate of the unknown values and error calculation associated** with the geostatistical  
4 linear interpolation, named “Kriging” (4). The Kriging is able to obtain a goodness of interpolation  
5 otherwise difficult to achieve with other techniques.

6 Regarding classical interpolation techniques, Kriging advantages are mainly:

- 7 a. Kriging estimation procedure is based on the real correlation law between the data  
8 (represented by the experimental semivariogram);  
9 b. Kriging interpolation leads to unique solutions;  
10 c. Kriging interpolates the data exactly (i.e. it returns the experimental value in the case  
11 of measured point estimation) and also it is able to give the estimated variance for  
12 each calculated value (in this way is possible to build both the phenomenon map and  
13 the estimation error map at once);  
14 d. Kriging results only depend from the model of semivariogram, from the distance  
15 between sampled points and unknown point to estimate, from sampling geometry and  
16 not from measured values.

17 For these reasons the kriging is defined as “best linear unbiased estimator” (B.L.U.EST.).

18 From a mathematical point of view Kriging estimation technique is defined by the equation 2:

$$20 \quad \hat{Z}_x = \sum_{i=1}^n \lambda_i(x) \cdot Z(x_i) \quad (2)$$

21  
22  
23 where  $\hat{Z}_x$  represents the unknown point value,  $Z(x_i)$  represents each measured valued  
24 which will be considered for the interpolation and  $\lambda_i(x)$  are appropriate weights which depend on  
25 the position and distance of each these point from unknown point and which satisfies the condition 3:  
26  
27

$$28 \quad \sum_{i=1}^n \lambda_i(x) = 1 \quad (3)$$

29  
30  
31 **5. results graphical restitution** (i.e. contour maps, block maps etc.) depending on the type of  
32 data and use of results. This procedure must be carried out after processing the data since the Kriging  
33 results are represented by a matrix of numerical values which is difficult to read by those who do not  
34 possess a specific know-how.  
35

## 36 DATA COLLECTION

37  
38  
39 This paragraph describes the phase of the experimental data collection and the construction of  
40 a datafile to develop the investigation. Our analysis only concentrated on horizontal curves in order to  
41 construct a prediction model where operating speed is expressed as a function of CCRs, R and both R  
42 and L (where L is the length of the horizontal).

43 In order to make the results of our work comparable with those proposed in international  
44 bibliography, the survey campaigns were carried out in correspondence to several Italian two lane  
45 rural road sections with the following characteristics (1) (2):

- 46  
47  
48  
49
- curves found in two lane rural roads;
  - lack of elements which could influence driver (such as working zones signals);
  - longitudinal grade below 5%;
  - low traffic flow;

- lane width between 3.00 and 3.75 m;
- shoulder width less than 1.50 m.

Twenty-two horizontal curves, on different Italian two lane rural roads, were chosen with a radius of 50 to 850 m.

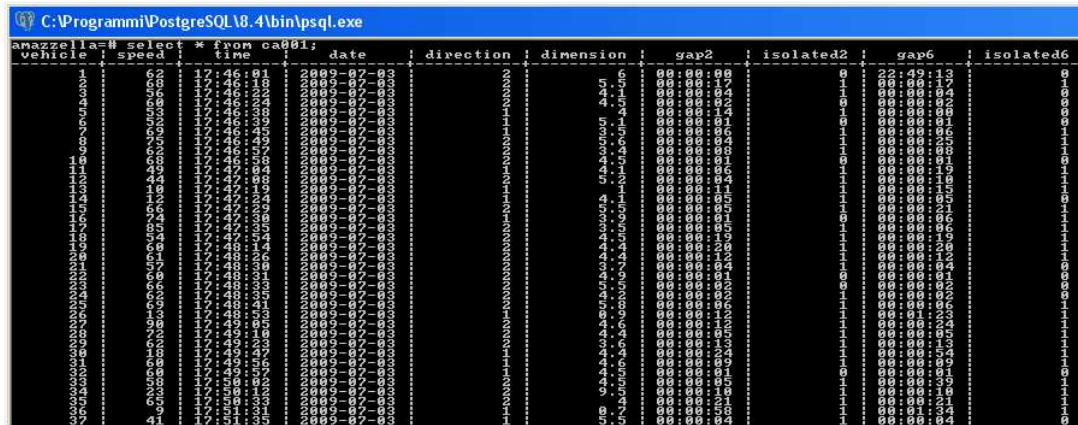
A radar instrument was used to reveal the instant speeds of the vehicles. It was located along the shoulder of each road section and hidden from the view of the drivers. This type of instrument was also chosen for its characteristic of being easy to hide from the view of the drivers and to install on a vertical mobile support at any site. In addition, the shape of the instrument is such that even if seen by the driver, it is not recognized as a survey or radar instrument. In this way it is possible do not influence driver behavior.

Speeds were registered in the center of each horizontal curve. The survey campaigns were carried out during the daytime, on dry pavement and in good weather conditions.

The radar also revealed real time speeds coming from both senses of direction, so for each section and for each vehicle, we recorded the length of the vehicle, the direction of flow, the instant speed, the date (day, month, year) and the time (the hour of the single passage).

The quantity of recorded data (about 1000 speeds for each curve) required an appropriate management system capable of processing these in order to make accurate analysis. For this reason, a database of Postgres was made to store each curve data (Figure 4). A table has been made for each curve section in which the following parameters were recorded: vehicle id, instant speed, time of recording, date of recording, direction of vehicle, length of vehicle and temporary gap between vehicles.

Necessary data to calculate  $V_{85}$  for each section of the curve surveyed were extracted by SQL queries to Postgres tables. This procedure allows to eliminate the vehicles that, even if travelling in opposite directions, crossed the recording spot less than 2 seconds. This allowed to eliminate any eventual recording errors, caused by the simultaneous passage of two vehicles (in opposite directions too). And secondly, by eliminating vehicles with gap 6 (going in the same direction less than 6 seconds) we were able to limit the study to isolated vehicles only. Isolated vehicles mean those with a travelling time distance of at least 6 seconds apart.



vehicle_id	speed	time	date	direction	dimension	gap2	isolated2	gap6	isolated6
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49
6	65.8	7:46:01	2009-07-09	300	1.4	00:00:00	0	22	49

FIGURE 4 A example of the used Postgres table

The decision to isolate vehicles was made in order to link vehicle speed only to the characteristics of the road and not other factors which could influence the speed, such as traffic flow. Besides, the study is only concentrated on cars, excluding trucks, commercial vehicles and motorcycles. The instant speed was recorded for over 300 vehicles on each curve. For these calculations only isolated cars (with 'gap2' of more than 2 seconds and 'gap6' with more than 6



seconds) were considered.

At this point the operative speed associated to each section of recording was calculated: the 85<sup>th</sup> percentile speed was calculated for probability of distribution for each curve. A data file was then constructed containing design values for each section of the curves (CCRs, R, length, DC) and operating speeds.

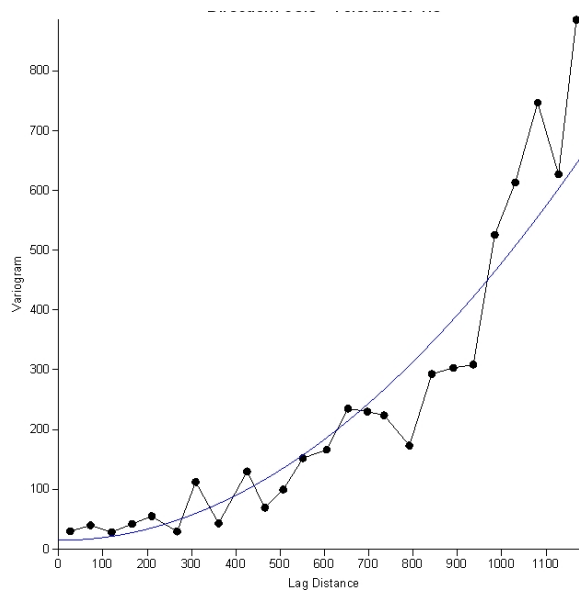
## GEOSTATISTICAL ANALYSIS OF COLLECTED SPEED DATA

The preliminary geostatistical study and successive application of the Kriging technique were conducted on imaginary grids composed of:

- a linear grid of points with coordinates (CCRs [gon/km],  $V_{85}$  [km/h]), for the construction of the model expressing CCRs vs  $V_{85}$ ;
- a linear grid of points with coordinates (R [m],  $V_{85}$  [km/h]), for the construction of the model expressing R vs  $V_{85}$ ;
- a 2D grid of points with coordinates (R [m], L [m],  $V_{85}$  [km/h]), for the construction of the model expressing R and L vs  $V_{85}$ .

### Model $V_{85}=f(\text{CCRs})$

The calculation of the data collected in the experimental semivariogram gave us the result in the Figure 5.



**FIGURE 5 Experimental semivariogram for  $V_{85}$  vs CCRs and its interpretation**

The modeling was conducted by choosing the best fitting theoretical model to describe the general trend and to further estimate variographic parameters using least square criterion.

The Figure 6 represents the Kriging result of the data carried out using parameters of variographic calculated models.

The trend of the model seems almost linear, like the other models suggested by international literature, but it has not a perfect linear trend. This demonstrates that the *a priori* choose of the regression type produces errors in all the phenomenon evaluation.

That is underlined also in the Figure 7, which shows the comparison between bibliographic

international models and our one.

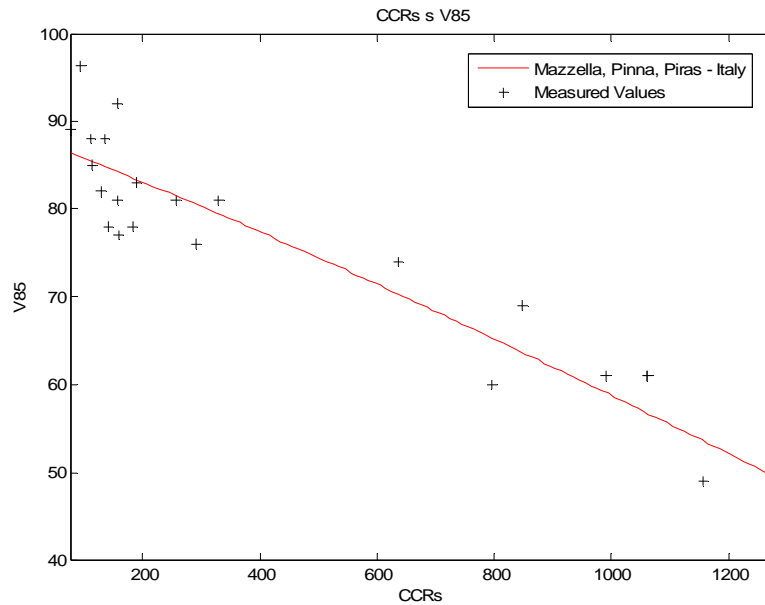


FIGURE 6 Kriging interpolation result for the  $V_{85}$  vs CCRs model

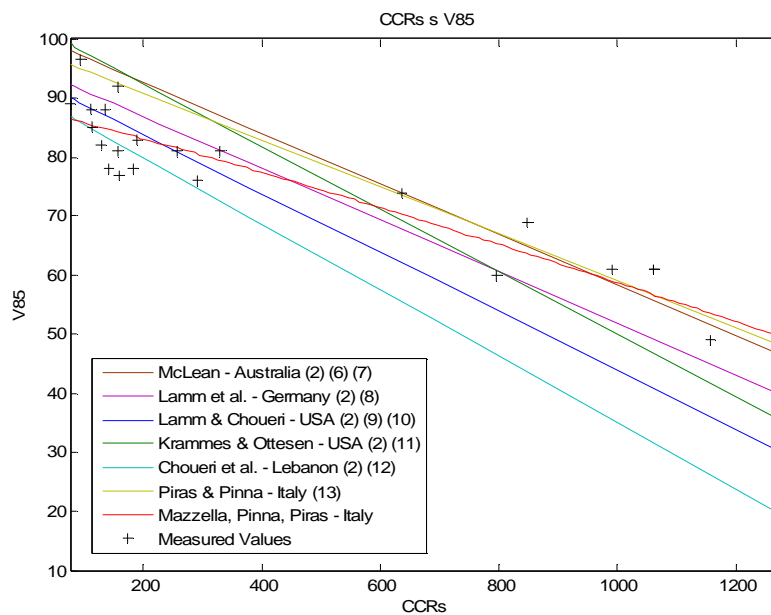
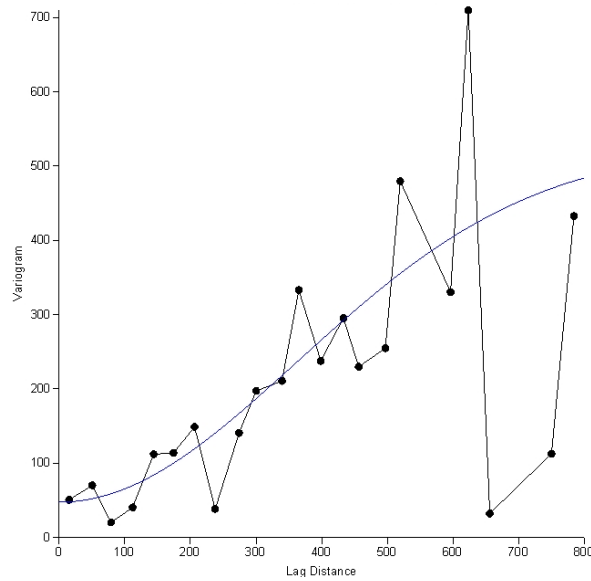


FIGURE 7 Comparison between international models and interpolation result for the  $V_{85}$  vs CCRs

The Figure 7 shows the evolution of speed on curves with a radius of 50 to 850 m. It shows a similar trend between the proposed model and most of those in use in various countries, even if the trend of our model, cutting the other ones, fits better the measured data.

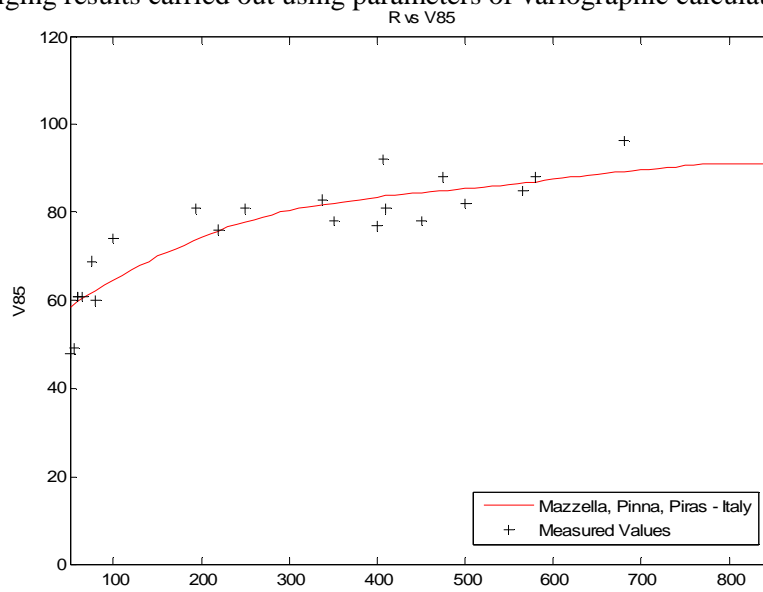
Model  $V_{85}=f(R)$

1 The calculation of the data collected in the experimental semivariogram gave us the result in  
 2 the Figure 8.



23  
 24 **FIGURE 8 Experimental semivariogram for V<sub>85</sub> vs R and its interpretation**

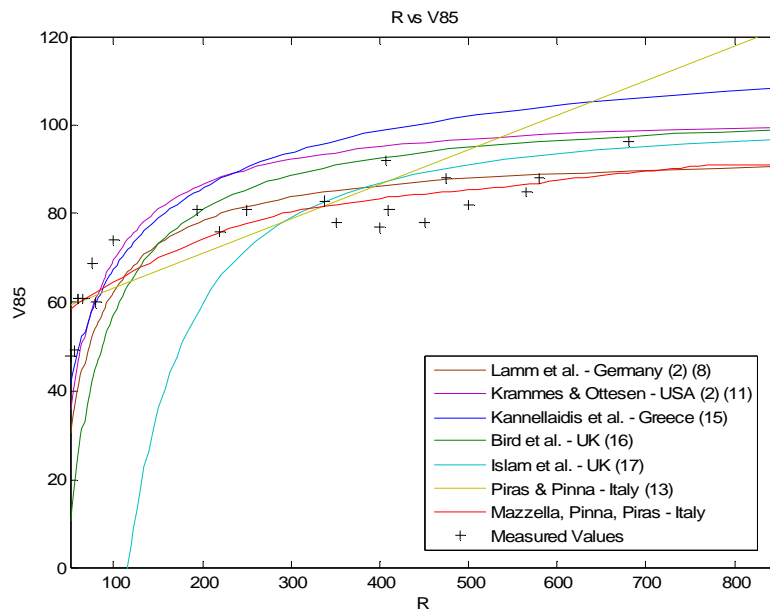
25  
 26 The modeling was conducted by selecting the best fitting theoretical model to describe the  
 27 general trend and to further estimate variographic parameters using least square criterion. The Figure 9  
 28 represents the Kriging results carried out using parameters of variographic calculated models.



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 48 **FIGURE 9 Kriging interpolation result for the V<sub>85</sub> vs R model**

49  
 50 The parabolic trend of the proposed model fits better than other models the distribution of  
 51 collected data. Also in this case this demonstrates that the *a priori* choose of the regression type  
 52 produces serious errors on the phenomenon evaluation.

1 Geostatistical curve built on measured data shows the existence of three different behaviors of  
 2 the phenomenon: first for radii between 50 m and 300 m, second for radii between 300 m and 750 m,  
 3 third for radii greater than 750 m (Figure 9). On this topic our research team is still working. Like the  
 4 previous case, the Figure 10 shows some the comparison between bibliographic international models  
 5 and our model.

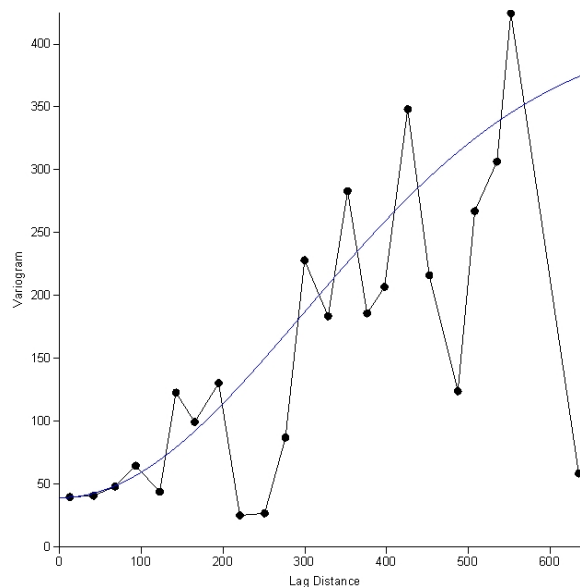


24 **FIGURE 10 Comparison between international models and interpolation result for the  $V_{85}$  vs  $R$**

26 The Figure 10 shows a similar trend between the proposed model and most of those in use in  
 27 various countries, especially for values of the radii between 150 and 700 m. Besides the difference  
 28 between our model (with the lowest) and the Kannellaidis one (with the greatest values) is almost  
 29 equal to 20 km/h.

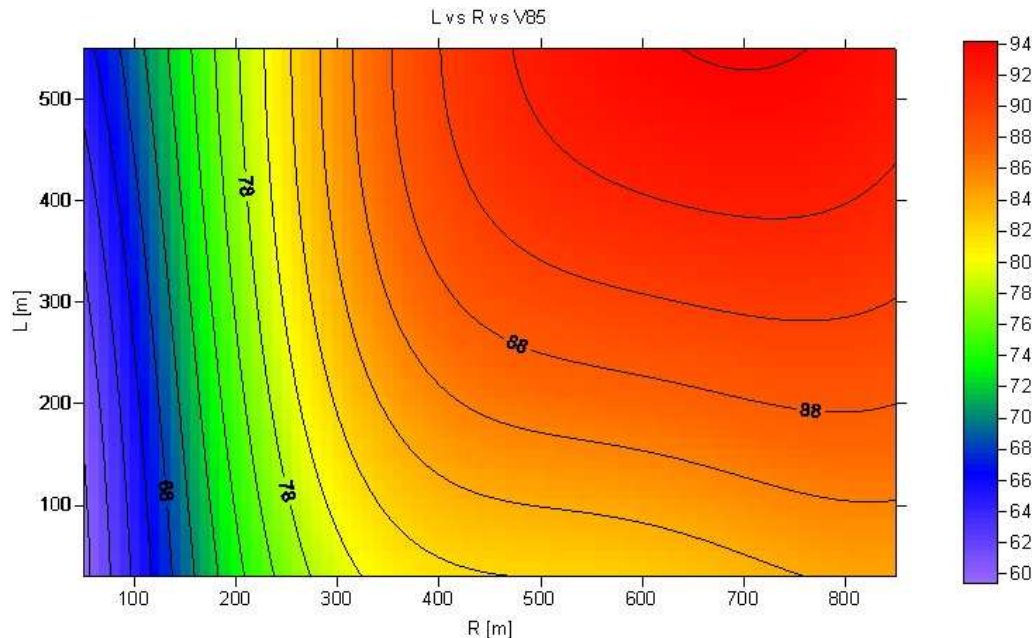
31 **Model  $V_{85}=f(R, L)$**

33 The calculation of the experimental semivariogram of the data collected provided the results  
 34 shown in the Figure 11.



1  
2  
3 **FIGURE 11 Experimental semivariogram for  $V_{85}$  vs R and L and its interpretation**  
4

5 The modeling was conducted by selecting the best fitting theoretical model to describe the  
6 general trend and to further estimate variographic parameters using least square criterion. The Figure  
7 12 represents the Kriging result of the data carried out using parameters of variographic calculated  
8 models.  
9



31 **FIGURE 12 Kriging interpolation result for the  $V_{85}$  vs R and L model**

32 The graph clearly shows the general  $V_{85}$  trend in variation to the radius R values and the  
33 length of the circular curves. In the image it is clearly shown the link between  $V_{85}$ , R and L.  
34

### 35 CONCLUSIONS

36  
37 There are numerous models in literature to calculate  $V_{85}$  in function of the rural road  
38 geometric characteristics, such as CCRs or horizontal curve radius. Almost all of these models were  
39 constructed using techniques own of Classical Statistics. No research group has still used  
40 Geostatistical analysis techniques even though they are largely used in many other fields of research  
41 inherent to Engineering sciences.

42 Some similarities can be seen when confronting the general trend of proposed models (using  
43 CCRs or R as independent variables) with those of respective models in use on an international level  
44 (created with classical Statistics techniques) (Figure 7 and 10). This clearly indicates a high  
45 implication of the validity of our proposed models, especially considering the fact that they are the  
46 first results using the geostatistical approach. In order to improve the models proposed in this paper  
47 we are still working on data collection on horizontal curves with radii greater than 700 m.

48 Using the correct study and modeling of such correlation, interpolations can be created which  
49 are based on the true variable of the phenomenon in exam where precise observations are obviously  
50 represented by exact demonstrations within the dominion of the study.

51 Geostatistics analytical instruments used correctly and in combination with the potentiality of

1 the Kriging algorithm, can create interpolations where the estimated variance is minimal (19).

2 In the current literature we find that many studies created models linking operating speed on  
3 curves to the horizontal curve features, or more precisely CCRs, radius of the horizontal curve, DC,  
4 etc. Even though it has been demonstrated that longitudinal grade or road alignment (tangent or curve)  
5 preceding the curve influence the speed in curves, there have been few studies that analyze such  
6 linking factors (2). For this reason, our research is leading towards the analysis of the relationship  
7 between driver behavior and other parameters, such as longitudinal grade, with the aid of geostatistical  
8 techniques. We are also in the process of defining a provisional 2D model where  $V_{85}$  is calculated by  
9 geostatistical analysis, contemporarily in function of R or CCRs and the longitudinal grade.

10 Furthermore, Geostatistics offers several analytical instruments that could be able to reveal the  
11 influence of the speed trend of vehicles going in the same direction for each section of study, meaning  
12 the element preceding the circular curve.

13 On the whole, the proposed work describes a new approach to study driver behavior in  
14 relation to the characteristics of the horizontal alignment, going beyond the limitations of classical  
15 statistics that:

- 16 1. chooses *a priori* regression type;
- 17 2. for the interpolation does not consider the true correlation between the sampled data.

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