

1 **LINKING RURAL ROAD ENVIRONMENT, SPEED AND SAFETY FACTORS WITH A**  
2 **‘TWO-STAGE’ MODEL: A FEASIBILITY STUDY**

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30

**1 ABSTRACT**

2 Speed on two-lane rural roads is a critical safety issue. In this regards various research perspectives  
3 have been adopted, including speed models (relating speed to design consistency factors) and safety  
4 models (which estimate safety using exposure data and design consistency variables). Unfortunately  
5 both approaches have often been carried out separately, and influences on speed choice have been  
6 limited only to geometrical variables. In contrast safety models are often expanded with wide array of  
7 exposure and risk factors.

8 The study aims to investigate the issue of speed and safety from a different perspective, using  
9 so called ‘two-stage’ model which estimates speed (using more explanatory variables) and further  
10 applies it in a simple safety performance function. This approach can be superior to the traditional  
11 approach as it preserves model parsimony while capturing the most important safety effects. The  
12 specific objective of the study is to prove feasibility of a ‘two-stage’ model in linking environment,  
13 speed and safety factors on a sample of Czech rural roads.

14 To this end, data collection was carried out on approx. 100 km (60 mi) of two-lane rural roads  
15 in the Czech Republic, using speed data from instrumented vehicle, manually collected road  
16 environment data, as well as crash and exposure data retrieved from national databases. Both models  
17 are developed, described and compared to the literature. It is concluded that approach is feasible, in  
18 spite of several current limitations. Planned further improvements and future practical applications are  
19 also listed.

20

## 1 INTRODUCTION

### 2 Operating speed models

3 Speed has been a critical issue within the traffic field; it was even described as one of the most  
4 important factors that road users consider in relation to convenience and efficiency of a certain route  
5 (1). Speed is also a key consideration in the geometric design and has a central role the road life cycle  
6 (2). At the same time speed has been recognized as the most influential risk factor (see 3, 4 or 5 for a  
7 review) – on Czech roads speeding (e.g. excessive and inappropriate speed) has been attributed to  
8 almost half of fatal road crashes, making it the most frequent cause of road deaths.

9 A host of road environment factors have been known to influence speed choice, including  
10 effects of road geometry, alignment, cross-section and roadside (6, 7, 8). A synthesis of Dutch studies  
11 (9) sums up also other road features which lead drivers to higher speed, such as presence of road  
12 marking or low density of alongside vegetation.

13 Within the field of traffic engineering, the concept of operating speed is of high importance.  
14 According to seminal TRB synthesis (1), the assessment of operating speeds enables to assess the  
15 expected speed changes of individual vehicles over successive road elements (tangents and curves).  
16 Inconsistency of operating speeds (for example differences between speed in tangent and curve) is  
17 regarded one of the symptoms that violates driver's expectation. Because consistent operating speeds  
18 are thought to be a product of consistent design, variables for evaluating design consistency are  
19 usually derived from operating speed (10, 11). With this focus large body of research has been devoted  
20 to modeling operating speed as a function of road parameters; among these horizontal curve radius or  
21 its transformations (inverse radius, degree of curve, curvature change rate, etc.) have been known as  
22 the most significant factor (1, 12, 13).

23 Therefore majority of models have included only horizontal curve radii or some of its  
24 derivatives; in spite of the fact that speed is very complex issue influenced by a number of other  
25 environmental variables, as mentioned in the previous text. For example the width (both road width  
26 and lane width) is regarded very influential cross-section parameter (14), however it has been included  
27 in few operating speed models only: a summary of North American operating speed studies (1, Table  
28 A-1) presents 23 models, of which 18 feature radius, while only 2 models consider road width.

29 It is worth noting that complete dependence of speed on road geometry is not always assured.  
30 For example Porter et al. (2) mention that geometric design decisions may not influence speeds unless  
31 very constrained dimensions are used.

### 32 Safety performance functions

33 Apart from the mentioned speed models, significant research efforts have also been devoted to  
34 development of so called safety performance functions (SPFs) or crash prediction models. These  
35 models provide relationships between crashes, exposure (e.g. traffic volume and segment length) and  
36 other potential risk factors (explanatory variables). Within focus of this study it is important that some  
37 SPFs also include variables related to speed, alignment or design consistency variables (for example  
38 10, 15 – 19). Consistency variables, used in these models, are usually computed through operating  
39 speed models, which were mentioned in the previous paragraph. Development of these SPFs thus  
40 involves two models – in the further text they will be referred to as *speed* models and *safety* models.  
41 Table 1 shortly summarizes variables, which were used in the mentioned SPFs.

1 **TABLE 1 Summary of Explanatory Variables Used in Several Speed and Safety Models**

Study	Explanatory variables <sup>1</sup>	
	Speed model	Safety model
Anderson et al. (10)	<i>R</i>	<i>AADT</i> , <i>L</i> , 4 consistency measures
Ng and Sayed (15)	<i>R</i>	<i>AADT</i> , <i>L</i> , <i>V</i>
Cafiso et al. (16)	<i>W</i> , <i>CCR</i>	<i>AADT</i> , 4 consistency and alignment measures
Camacho-Torregrosa et al. (17)	<i>R</i>	<i>V</i>
de Oña et al. (18)	<i>R</i>	<i>AADT</i> , <i>L</i> , <i>V</i>
Montella and Imbriani (19)	<i>R</i> , <i>CCR</i> , <i>G</i> , <i>L<sub>T</sub></i>	<i>AADT</i> , up to 5 consistency and alignment measures

2 <sup>1</sup> Abbreviations: *R* – radius; *W* – width; *CCR* – curvature change rate; *G* – vertical grade; *L<sub>T</sub>* – length of  
3 preceding tangent; *AADT* – traffic volume; *L* – length; *V* – speed consistency measure.

4 From Table 1 it is evident that some speed models rely on explanatory power of a single variable  
5 (horizontal curve radius). In contrast some of safety models introduce a large number of variables;  
6 some even applied the same variables in both models at the same time.

### 7 ‘Two-Stage’ model

8 The above review on speed and safety models demonstrate that while speed models are usually kept  
9 parsimonious, safety models may be relatively complex. In this regards an interesting approach has  
10 been recently applied by Chen et al. (20) in studying interrelationships of geometry, speed and safety  
11 on roundabouts. They developed a ‘two-stage’ model – firstly approach speed was modeled, which is  
12 applied in a crash prediction model in the second stage. According to the authors, such an approach  
13 can be superior to the SPFs directly containing design variables as it preserves model parsimony while  
14 capturing the important safety effects of design changes. It means that, as opposed to the models in the  
15 previous paragraphs, speed model (stage 1) include more variables, while safety model (stage 2) is  
16 parsimonious.

17 The objective of this study is to prove feasibility of development and application of a  
18 combination of speed and safety models (in so called ‘two-stage’ model) in the study of environment,  
19 speed and safety factors on a sample of Czech rural roads. The two stages will be as follows:

- 20 1. Using road environment factors to estimate speed on a segment (speed model). Estimated  
21 speeds on individual tangents and curves will be used to compute indicator of speed  
22 consistency (as a difference between the speeds in tangent and curve).
- 23 2. Speed consistency indicator will be used as additional explanatory variable in safety  
24 performance function (safety model) in order to predict expected crash frequency.

25 Compared to previous applications of SPFs with design consistency variables (Table 1), the proposed  
26 approach will have some novel features.

- 27 – As opposed to simple speed models (often based on one variable only), speed model will  
28 consider influence of more variables, which are known to impact driving speed choice.  
29 Curvature change rate (*CCR*) will be used as one of explanatory factors, since is seen as the  
30 most successful parameter in explaining much of the variability in operating speeds (21); other  
31 factors will consider not only road geometry or alignment, but rather broader road  
32 environment. Different speed models will be developed separately for tangents and for curves.
- 33 – In contrast, safety model will be developed as much as simple as possible. The objective is to  
34 achieve parsimony (in a number of variables and one function form common for both tangents  
35 and curves), as opposed to some complex functions illustrated in Table 1, some of which are  
36 further distinguished between tangents and curves.

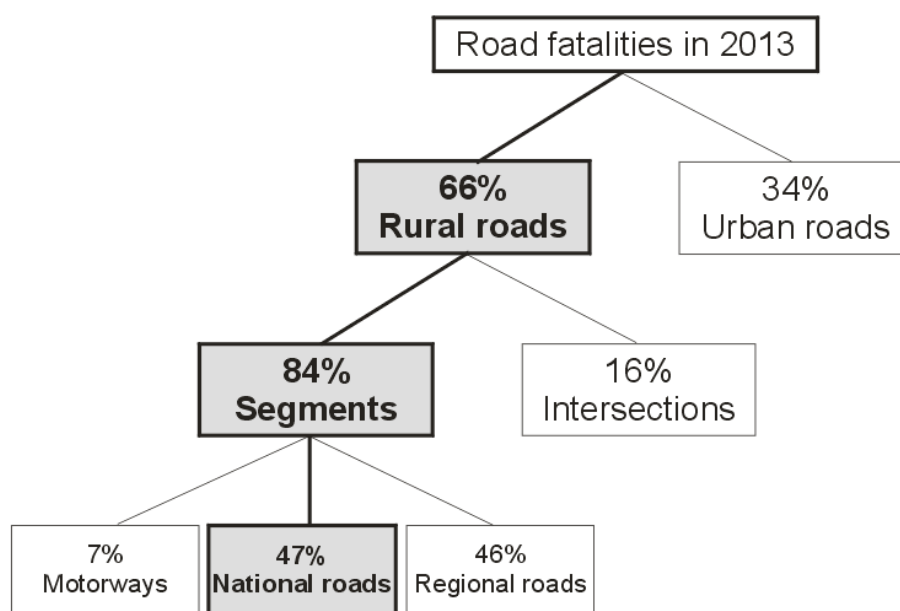
1 The following section 2 describes data collection and characteristics and methods used to  
 2 develop the ‘two-stage’ model. Results are reported in section 3, together with descriptions of several  
 3 comparisons. The final section 4 brings discussions and conclusions, aiming to assess the achieved  
 4 feasibility, further improvements and potential practical applications.

5

## 6 2 DATA AND METHODS

### 7 Study location

8 The authors’ intention was to apply the introduced concept on a sample, which will be representative  
 9 of the most critical settings within Czech road network. To this end, disaggregated Czech Traffic  
 10 Police data were studied. Figure 1 provides a division of Czech road fatalities counts by road settings  
 11 (rural or urban roads), road network elements (segments or intersections) and their categories:  
 12 motorways, national roads (1<sup>st</sup> class roads) or regional roads (2<sup>nd</sup> and 3<sup>rd</sup> class roads).



13

14 **FIGURE 1 Division of Czech road fatalities counts by road settings according to 2013 police**  
 15 **data.**

16 Each level of graph provides several blocks describes in terms of percentages of road fatalities in  
 17 2013. The most critical settings (in grey blocks) is rural segments of national roads. Specifically in this  
 18 category, approximately 40% of fatalities are related to curve crashes; within these crashes on national  
 19 roads speeding was attributed as the main cause of almost 40% of fatalities.

20 Given this focus, the study sample was chosen in one of the Czech regions (Kraj Vysočina).  
 21 Of 5 national roads in this region, the two roads (No. 19 and 34) with the highest traffic volumes and  
 22 risk were selected for the study. The roads are paved, two-lane, undivided, approximately 7 meters (23  
 23 ft) wide. Approximate traffic volume (AADT) is between 5,000 and 10,000 vehicles and general speed  
 24 limit is 90 km/h (54 mph). After excluding the road sections in built-up areas (through-roads), their  
 25 total length was approximately 100 km (60 mi).

## 1 Speed and alignment data

2 The roads in question were driven through in two weekdays in November 2013, in one direction, as  
3 close as possible to free-flow speed. The inspection vehicle of CDV – Transport Research Centre  
4 (Centrum dopravního výzkumu, v.v.i.) was used for this purpose, instrumented with several position  
5 sensors (gyroscope, accelerometer, odometer) as well as controller area network (CAN) bus, whose  
6 data are synchronized and positioned using a precise GPS with the frequency of 10 Hz. At the typical  
7 speed of 90 km/h (54 mph) the speed synchronization period equals to 2.5 m (8.25 ft) of driven  
8 distance.

9 In the Czech Republic it is difficult to obtain periodically updated and precise road design  
10 plans. Thus a method had to be developed in order to obtain alignment parameters and conduct  
11 segmentation into tangents and curves. The development and the pilot (non-automated) application of  
12 the methodology is described elsewhere (22); for this study it was programmed into an in-house web  
13 module in order to ensure its semi-automation and wider application. The employed calculation  
14 procedure consisted of several steps:

- 15 – Transformation of GPS data points into the Czech planar coordinate system JTSK.
- 16 – Calculation of distances and angles between points in order to calculate radii and lengths for  
17 each three consecutive points.
- 18 – Calculation of curvature change rate (CCR).
- 19 – Segmentation of data points into tangent and curve sections using CCR threshold; based on  
20 several sensitivity tests, its value was set at 80 gon/km. The process resulted in 316 segments:  
21 158 tangents and 158 curves.

22 In terms of design consistency, each segment may be characterized by its values of speed and  
23 CCR. Given the sampling frequency of used GPS technology, both speed and CCR change  
24 continuously within segments. In order to smooth out the speed and CCR values 85<sup>th</sup> percentile of  
25 speed ( $V_{85}$ ) and 85<sup>th</sup> percentile of CCR ( $CCR_{85}$ ) were determined for each segment.

## 26 Road environment data

27 Based on the review of speed factors in the first section of the paper, several of them were chosen as  
28 candidate explanatory variables. To this end, data on following variables for each segment were  
29 manually collected using Google Maps (with in-house web environment based on Street View,  
30 developed with Google Maps API) and categorized as follows:

- 31 – *Roadside vegetation*: none or bushes; single trees; trees in a row or forest.
- 32 – *Road marking* (separation of driving directions): no line or broken line; solid line.
- 33 – *Delineator posts*: absent; present.
- 34 – *Guardrails*: absent; present.
- 35 – *Vertical grade*: absent (flat); present (slope).
- 36 – *Roadway width*. This data were extracted from road database data maintained by Czech Road  
37 and Motorway Directorate and assigned to individual segments. Where values were changing  
38 within a segment, an average value was used. Afterwards it was categorized, based on Czech  
39 road width categories (7.5, 9.5, 11.5 meters; i.e. approx. 25, 31 and 38 ft) into following four  
40 width classes: 7.5 m or less; 7.6 – 9.5 m; 9.6 – 11.5 m; 11.6 m or more.

41

## 1 Crash and exposure data

2 In studies related to driving behavior, speed and alignment, various authors use different crash types  
 3 for their analyses. For example Anderson et al. (10) used both single- and multi-vehicle crashes but  
 4 excluded, among others, crashes with animals. Turner et al. (23) combined loss-of-control and head-on  
 5 crashes; as well as Dietze and Weller (24) who used only single-vehicle and overtaking crashes. In this  
 6 study all single-vehicle crashes were used. In total 5 years (2009 – 2013) were considered, taking into  
 7 account crashes of all severity levels (property damage only, slight/severe/fatal injury). Using this  
 8 definition, georeferenced crashes on selected roads were retrieved from the Police database and  
 9 assigned as a crash frequency to each segment.

10 As a risk exposure indicator, traffic volume data (AADT) were retrieved from the National  
 11 Traffic Census data of the Czech Road and Motorway Directorate. For curves, length of preceding  
 12 tangent was added as another factor. Descriptive statistics for all the mentioned data are listed in  
 13 following Tables 2 and 3, separately for curves and tangents.

14 **TABLE 2 Descriptive Statistics of Collected Data (Scale Variables)**

	Variables <sup>1</sup>	<i>N</i>	<i>AADT</i> [veh/day]	<i>L</i> [m] <sup>2</sup>	<i>CCR</i> <sub>85</sub> [gon/km]	<i>V</i> <sub>85</sub> [km/h]	<i>L<sub>pre</sub></i> [m]
Curves	Min.	0	1,122	21	19,75	48.01	0
	Max.	4	12,096	2,403	984.77	95.55	2,924
	Mean	0.31	5,050.43	188.94	216.82	77.84	406.41
	SD	0.69	2,966.09	269.76	162.39	11.03	519.90
Tangents	Min.	0	1,122	30	17.77	48.17	
	Max.	10	12,096	2,924	177.01	111.67	
	Mean	0.78	5,061.64	423.24	60.70	79.02	
	SD	1.66	3,023.87	541.07	23.33	11.30	

15 <sup>1</sup> Abbreviations: *N* – 5-year frequency of single-vehicle crashes; *AADT* – traffic volume; *L* – length; *CCR*<sub>85</sub> –  
 16 85<sup>th</sup> percentile of curvature change rate; *V*<sub>85</sub> – 85<sup>th</sup> percentile of speed; *L<sub>pre</sub>* – preceding tangent length (for  
 17 curves only)

18 <sup>2</sup> Units: 1 m (meter) = 3.3 ft; 1 km (kilometer) = 0.6 mi; 1 gon = 10/9°

19

1 **TABLE 3 Descriptive Statistics of Collected Data (Categorical Variables)**

Variable	Categories	Curves		Tangents	
		Freq.	%	Freq.	%
Roadway width	1 (up to 7.5 m)	66	41.77	67	42.41
	2 (7.6 – 9.5 m)	19	12.03	20	12.66
	3 (9.6 – 11.5 m)	60	37.97	60	37.97
	4 (11.5 m +)	13	8.23	11	6.96
Vegetation	0 (none / bushes)	25	15.82	27	17.09
	1 (single trees)	58	36.71	52	32.91
	2 (trees / forest)	75	47.47	79	50.00
Road marking	0 (none / broken)	77	48.73	44	27.85
	1 (solid)	81	51.27	114	72.15
Delineator posts	0 (absent)	15	9.49	10	6.33
	1 (present)	143	90.51	148	93.67
Guardrails	0 (absent)	116	73.42	116	73.42
	1 (present)	42	26.58	42	26.58
Vertical grade	0 (flat)	115	72.78	92	58.23
	1 (slope)	43	27.22	66	41.77

2

3 The collected data may be grouped as following variables, according to the two stages of modeling  
4 (see Table 4):

5 **TABLE 4 Overview of Variables Used in Speed and Safety Models**

	Explanatory variables	Response variable
Speed model	Environment factors	Speed
	Length of preceding tangent (for curves only)	
	Curvature change rate	
Safety model	Exposure (length and AADT)	Crash frequency
	Speed consistency	

6 Note that speed models will be developed separately for tangents and curves, while safety model will  
7 be one for both tangents and curves. Both models will be developed using generalized linear modeling  
8 feature in SPSS (procedure GENLIN). Segment number is referred to as  $i$ . Speed model was  
9 considered in a following form:

$$10 \quad V_{85,i} = \beta_0 + \sum_{i=1}^n \beta_i \cdot x_i \quad (1)$$

11 where:

12  $V_{85,i}$  ... speed

13  $x_i$  ... explanatory variables (from Table 3)

14  $\beta_i$  ... regression coefficients to be estimated



1 For safety model negative binomial distribution with logarithmic link function was used. The model  
2 has following form:

$$3 \quad P_i = \exp(\beta_0) \cdot AADT_i^{\beta_1} \cdot L_i^{\beta_2} \cdot \exp(\beta_3 \cdot |\Delta V_{85,i}|) \quad (2)$$

4 where:

- 5  $P_i$  ... expected crash frequency  
6  $AADT_i$  ... traffic volume [veh/day]  
7  $L_i$  ... segment length [km]  
8  $|\Delta V_{85,i}|$  ... speed consistency indicator [km/h]  
9  $\beta_i$  ... regression coefficients to be estimated

10 Speed consistency was quantified as absolute difference of 85<sup>th</sup> percentile speeds ( $|\Delta V_{85}|$ ) between  
11 curve  $i$  and tangent  $i + 1$ .

12

### 13 3 RESULTS

#### 14 Stage 1: Speed model

15 The results related to speed model (Stage 1) are listed in Table 5. Only variables with statistical  
16 significance at 5% level ( $p \leq 0.05$ ) are reported; two of original variables were not significant at this  
17 level (presence of delineator posts and preceding tangent length).

18 **TABLE 5 Regression Coefficients and Statistical Significance of Explanatory Variables in Speed**  
19 **Models**

Variable	Category	Regression coefficients		<i>p</i> -values*		Interpretation
		Curves	Tangents	Curves	Tangents	
Roadway width	1	0	0	0.002	0.000	Positive
	2	0.50	3.12			
	3	-0.36	4.26			
	4	9.78	14.01			
Vegetation	0	0	0	0.011	0.050	Positive
	1	-0.56	1.56			
	2	4.01	4.80			
Road marking	0	–	0	n.s.	0.002	Positive
	1	–	5.37			
Guardrails	0	0	0	0.015	0.005	Positive
	1	4.05	4.89			
Vertical grade	0	0	0	0.023	0.052	Negative
	1	-3.62	-3.03			
$CCR_{85}$	–	-0.03	-0.07	0.000	0.031	Negative

20 \* n.s. – not significant

21

1 Interpretation in the last column is based on signs of regression coefficients (i.e. directions of  
2 relationships between predictor and predicted variable):

- 3 – Increase of road width is associated with higher speed. This is generally in line with literature  
4 (14, 25, 26).
- 5 – Speed increases with increasing density of vegetation, which is not completely consistent with  
6 past knowledge (9). However it was reported that ‘continuous wall of overgrowth’ may lead to  
7 speed increase, compared to bushes along the road, which may decrease speed, (25).
- 8 – Presence of solid line is associated with higher speeds in tangents. This may be consistent with  
9 a finding fact that presence of road marking (in contrast to no marking) is related to higher  
10 speed (9).
- 11 – Higher speeds in segments with guardrails also confirm previous findings studies (9).
- 12 – Both vertical grade and increase of curvature change rate are associated with decrease of  
13 speed, which is consistent with previous studies (1, 25).

#### 14 Stage 2: Safety model

15 The results related to safety model (Stage 1) are listed in Table 6. Statistical significance at 5% level ( $p$   
16  $\leq 0.05$ ) was reached for AADT and length, but not for speed consistency, which achieved statistical  
17 significance at 13.4% level. Nevertheless due to its presumed causal role, it was kept in the model.

18 **TABLE 6 Regression Coefficients and Statistical Significance of Explanatory Variables in Safety**  
19 **Model**

Parameter	Regression coefficients	95% confidence interval boundaries	$p$ -values
(Intercept)	-6.725	-9.427 -4.023	0.000
$\ln(AADT)$	0.838	0.523 1.154	0.000
$\ln(L)$	0.941	0.778 1.104	0.000
$ \Delta V_{85} $	0.030	-0.009 0.068	<b>0.134</b>

20 Signs of all regression coefficients are positive. The exception of  $|\Delta V_{85}|$  – although majority of its  
21 confidence interval is in positive values, it is also partly below zero. However overall the regression  
22 coefficients are positive, i.e. increasing values of predictors are associated with higher crash  
23 frequencies. This confirms expectations based on general knowledge from crash prediction modeling.

24 In addition sizes of individual regression coefficients may be compared to past studies, which  
25 used the same model form. Table 7 compares the above results with 4 studies – the values are not  
26 identical, but mostly relatively close. The consistent values are in bold text: for  $\ln(AADT)$  between 0.6  
27 and 0.9, for  $\ln(L)$  between 0.8 and 1.0, for  $|\Delta V_{85}|$  between 0.02 and 0.05. Potential inconsistencies  
28 may arise from variable conditions between individual studies: for example different crash types or  
29 incomparable ranges of explanatory variables.

30 **TABLE 7 Comparison of Regression Coefficients with Several Other Studies**

Variable	This study	Fitzpatrick et al. (27, equation 35)	Ng and Sayed (15, model 2)	Dietze and Weller (24, Table 9)	de Oña et al. (18, Table 8, average)
$\ln(AADT)$	<b>0.838</b>	<b>0.922</b>	<b>0.585</b>	<b>0.585</b>	1.129
$\ln(L)$	<b>0.941</b>	<b>0.842</b>	<b>0.887</b>	0.216	<b>0.955</b>
$ \Delta V_{85} $	<b>0.030</b>	0.066	<b>0.048</b>	<b>0.033</b>	<b>0.015</b>

1 Overall performance of safety model may be quantified in terms of various indicators, for example  
 2 Akaike information criterion (AIC), overdispersion parameter or ‘proportion of systematic variation in  
 3 the original crash dataset explained by the model’ or %SV (28). Contribution to model’s explanatory  
 4 power, which is caused by inclusion of additional variable, then equals the difference in values of  
 5 these indicators. Both original model (with AADT and length) and full model (with AADT, length and  
 6  $|\Delta V_{85}|$ ) were compared this way. The results are listed in Table 8.

7 **TABLE 8 Comparison of Performance of Original and Full Models**

Model variant	This study <sup>1</sup>			Anderson et al. (10, model 2)
	AIC	O.d.p.	%SV	%SV
Original model	480.682	0.281	91.306	65.86
Full model	480.526	0.260	91.956	66.51
Difference	-0.156	-0.021	+0.650	+0.66

8 <sup>1</sup> Abbreviations: AIC – Akaike information criterion; O.d.p. – overdispersion parameter; %SV – proportion of  
 9 systematic variation explained.

10 The differences in all indicators are small. Nevertheless such small contributions have also been  
 11 reported in other similar studies (e.g. 10), as reported in the last column of Table 8; therefore  $|\Delta V_{85}|$   
 12 was kept in a model.

13 One potential application of model results may be in network screening, i.e. identification of  
 14 hazardous road segments. Traditionally estimates from crash prediction models are used for this  
 15 purpose; should we prove valid relationship of  $|\Delta V_{85}|$  to safety, speed consistency may be used as a  
 16 surrogate measure in a proactive way (before occurrence of crashes). Preliminary results from the  
 17 same road sample, but using observed speeds and limited to curves only, were reported in a recent  
 18 paper (29). Both trends were found to be relatively similar, with Pearson correlation coefficient  
 19 between ranking both assessments equal to 0.62 in the preliminary study, and improved to 0.68 while  
 20 using modeled speeds and both tangents and curves.

21

## 22 4 DISCUSSION AND CONCLUSIONS

23 Speed on two-lane rural roads is a critical safety issue. Within traffic and safety engineering, speed has  
 24 been studied from several different perspectives. Two of them are referred to as speed models (relating  
 25 speed to design consistency factors) and safety models (where safety is estimated using exposure data,  
 26 also enriched by design consistency variables).

27 However research in both domains has often been carried out separately; their combination  
 28 would be useful (2). What is more, although it is well known that a host of road environment factors  
 29 influences speed choice, speed models have usually employed only few selected variables. In contrast,  
 30 safety models often use wide array of exposure and risk factors. At the same time in a different field –  
 31 studying roundabout safety – an opposite approach has been applied. So called ‘two-stage’ model  
 32 estimates speed (using more explanatory variables) which is further applied in a relatively simple  
 33 safety performance function (SPF). According to the authors (20), such an approach can be superior to  
 34 the SPFs directly containing design variables as it preserves model parsimony while capturing the  
 35 important safety effects.

1 The objective of this study was to prove feasibility of development and application of a  
2 combination of speed and safety models (in a ‘two-stage’ model) in the study of environment, speed  
3 and safety factors on a sample of Czech rural roads. To this end, data collection was carried out on  
4 approx. 100 km of two-lane rural roads in one of Czech regions. Using instrumented vehicle, speed  
5 and alignment data were obtained – 316 segments were created (158 tangents and 158 curves). These  
6 segments were assigned road environment data, based on their expected relation to speed choice  
7 (roadside vegetation, road marking, delineation, guardrails, vertical grade, roadway width), as well as  
8 single-vehicle crash frequency and exposure data (AADT and length). Data were used to develop and  
9 study speed and safety models in the following steps:

- 10 – Speed model utilized 8 potential explanatory variables; all but 2 of them were statistically  
11 significant at 5% level. All these variables had expected direction of relationship to speed.
- 12 – Modeled speeds were used to compute indicator of speed consistency  $|\Delta V_{85}|$ .
- 13 – Safety model was developed using AADT, length and speed consistency. 5% statistical  
14 significance was achieved for AADT and length, but not for speed consistency, which  
15 achieved statistical significance at 13.4% level. Nevertheless due to its presumed causal role,  
16 it was kept in the model. Sizes of regression coefficients were compared to 4 past studies and  
17 found to be in relative agreement. Also overall model performance was tested; however  
18 improvements due to added speed consistency were small, which was found also in other  
19 studies.

20 To sum up, final results are maybe less relevant than expected. Significances of variable influence and  
21 model performance could be higher. Nevertheless the findings seem plausible and are mostly  
22 comparable to other similar studies. Differences may results from the following limitations of the  
23 presented study:

- 24 – *Road environment data.* There are other variables, which were not used in this study, and  
25 could potentially improve the quality of speed models, such as pavement quality or  
26 superelevation in curves. Vertical grade could be used in a more quantitative way. In addition  
27 road environment data could be registered while riding, rather than from Google Maps, as was  
28 done in this study.
- 29 – *Speed data collection.* Data were collected within a single drive only in a single direction.  
30 Although there was an attempt to adapt driving speed to the free-flow as close as possible, the  
31 collected data may not be representative of the driving population and are considered as an  
32 approximation of operating speed only. In this regard, more drives, possibly with more  
33 drivers, could offer more reliable data, leading to different results.
- 34 – *Speed consistency measure.* Several studies recommended not to rely on a simple indicator of  
35  $|\Delta V_{85}|$ , which was used in the presented study, since it may underestimate the real speed  
36 reduction (30). Other measures, such as 85<sup>th</sup> percentile of maximum speed reduction, may  
37 circumvent the issue (31). Speed may be also collected in several points of speed profile, from  
38 the approach tangent through the curve and departure tangent (1).
- 39 – *Crash sample.* The sample of modeled crashes was generally small, and further reduced by  
40 using single-vehicle crashes only. Low numbers may influence quality of safety modeling, due  
41 to low statistical significances and also a ‘low mean problem’ which biases estimate of  
42 overdispersion (32).

43 These limitations are being addressed in further stages of the projects: study sample is enlarged in both  
44 time and space, using vehicle fleet speed data from repeated drives in both directions (33). An  
45 improvement of evaluation methodology is planned as well, considering more data collection spots in

1 curves and their surroundings. It will enable improving the study quality and testing the validity of the  
2 findings presented in this feasibility study, as well as comparison to other studies.

3 Future practical applications of an improved concept may include proactive network screening  
4 (i.e. identification of hazardous road segments) without having to rely on crash occurrence only. Once  
5 hazardous segments are identified, potential countermeasures may be proposed. Common  
6 countermeasures in horizontal curves include warning signs, road marking, recommended speeds or  
7 reduced speed limits (34). A number of them have been recommended as safety-beneficial and low-  
8 cost (35). Joint analysis of environment, speed and safety factors, as proposed in this paper, will help  
9 decide on the most suitable countermeasures in order to tackle the issue of Czech rural roads safety.

10

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