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Additional Information

1 **ANALYSIS OF THE INFLUENCE OF GEOMETRIC DESIGN CONSISTENCY ON**
2 **VEHICLE CO₂ EMISSIONS**

3
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31 ABSTRACT

32 Highway vehicles driving on rural roads account for more than 50% of all CO₂ emissions produced
33 by the transportation sector in Europe. Although the policy measures to mitigate Greenhouse Gas
34 emissions are increasing, these do not include policies aimed at reducing emissions by means of
35 highway geometric design, which significantly influences drivers' speeds and accelerations and,
36 consequently, plays a major role on fuel consumption and emissions.

37 Therefore, the main objective of this research is to study the influence of the geometric
38 design consistency on vehicle CO₂ emissions. To do this, continuous speed data were collected on
39 47 homogeneous road segments by means of Global Positioning System devices. Vehicle CO₂
40 emissions were estimated by applying the VT-micro model, whereas geometric design consistency
41 was assessed considering different global consistency models.

42 As a conclusion, vehicle CO₂ emissions decreases as the consistency level of a
43 homogeneous road segment increases. Specifically, a good consistency road segment has been
44 found to present an emission rate 20-30% lower than a poor-consistent one. Therefore, the design
45 of consistent roads allows, in addition to maximize road safety, to help to achieve more
46 environmentally sustainable highways, reducing CO₂ emission production.

47

48 *Keywords:* geometric design consistency, CO₂ emission, two-lane rural road, traffic operation,
49 environmentally-friendly transport, naturalistic data

50

51 **1. INTRODUCTION**

52 **1.1. Transportation and Greenhouse Gas emissions**

53 In 2015, the transportation sector contributed 25.8% of total EU-28 Greenhouse Gas (GHG)
54 emissions (European Environment Agency, 2017). These emissions were 23% above 1990 levels,
55 presenting a general increasing trend despite a decline between 2008 and 2013. Specifically,
56 highway vehicles driving on rural roads accounted for more than 50% of all CO₂ emissions
57 produced by the transportation sector (European Environment Agency, 2017).

58 According to the United States Environmental Protection Agency (2016), emissions from
59 transportation also rose significantly (17%) in this country between 1990 and 2014. Additionally,
60 the combustion of fossil fuels associated to transport accounts for about 30% of total US CO₂,
61 being the second largest source of CO₂ emissions in 2014. The largest sources were associated to
62 passenger cars (42.4%), medium- and heavy-duty trucks (23.1%), and light-duty trucks (17.8%)
63 (US EPA, 2016).

64 Faced with this situation, both the United States and Europe have recently proposed some
65 strategies to deal with issues regarding air pollution from transport. These policies mainly focus
66 on intelligent transport systems, eco-driving courses, urban mobility/smart cities, road pricing and
67 internalization, and fuel switching and efficiency (European Commission, 2016; IPCC, 2015).
68 However, these policies do not include measures related to highway geometric design despite the
69 fact that the geometric alignment can meaningfully influence vehicle fuel consumption and
70 emissions.

71 Some researchers have studied the influence of geometric design on fuel consumption and
72 emissions, being the longitudinal grade the most examined geometric variable. Boriboonsomsin
73 and Barth (2009) analyzed the fuel consumption from a single vehicle (2007 Nisan Altima)

74 travelling at 90 km/h along three routes: (i) level, (ii) upgrade, and (iii) downgrade. Both upgrade
75 and downgrade routes had an average grade about 4% and each route was covered three times. The
76 findings showed a 2nd-degree relationship between fuel consumption and longitudinal grade ($R^2 =$
77 0.93). Specifically, a great influence of this geometric variable on fuel economy of light-duty
78 vehicles was identified. In addition, the vehicle fuel economy on level routes was 15-20% larger
79 than for upgrade and downgrade routes.

80 Other studies have used microscopic and dynamic models to obtain speeds and
81 accelerations, which are easier and less expensive than field data collection (Park and Rakha, 2006;
82 Ko et al., 2012; Ko et al., 2013).

83 Regarding the influence of horizontal alignment on GHG emissions, Llopis-Castelló et al.
84 (2018a) concluded that CO₂ emission rates on an entire road segment increase with the Curvature
85 Change Rate (*CCR*). Low *CCR* values mean that the homogeneous road segment is mainly
86 composed of flat curves and long tangents, allowing drivers to reach greater speeds without large
87 speed variations. By contrast, homogeneous road segments with high *CCR* values impose
88 important geometric controls on drivers, leading to lower speeds and greater speed variations. In
89 this way, CO₂ emission rate was lower for higher average speeds and lower speed variations. For
90 LDV5 vehicles, up to 13% of emissions can be attributed to the horizontal alignment. This is a
91 congruent conclusion to the findings by Boriboonsomsin and Barth (2009).

92 This example is only related to operational emissions, i.e. emissions due to road lifecycle
93 and vehicle manufacturing have not been considered. In addition, this does not mean that 13% of
94 the emissions could be reduced by optimizing the horizontal alignment: there are many constraints
95 that prevent designers from creating a minimum-impact horizontal alignment. However, some sort
96 of emission reduction could be addressed having this into account in the alignment optimization

97 process.

98 **1.2. Geometric design consistency**

99 Geometric design consistency is closely related to road safety and can be defined as how drivers'
100 expectations and road behavior relate. Thus, a consistent road aims to provide a harmonious
101 driving experience free of surprises, whereas an inconsistent road segment tends to show up
102 numerous unexpected events to drivers, which usually results in wrong driver decisions increasing
103 the likelihood of crash occurrence.

104 The most commonly method to assess geometric design consistency relies on the
105 examination of the operating speed profile (Gibreel et al., 1999). Operating speed is usually
106 defined as the 85th percentile of the speed distribution for passenger cars under free-flow conditions
107 and favorable weather conditions (V_{85}).

108 There are two types of consistency models: local and global. Local models aim at
109 identifying localized issues, such as great differences between design and operating speeds or
110 sudden speed reductions; whereas global models analyze the overall speed variation through an
111 entire homogeneous road segment, allowing the estimation of the number of road crashes by means
112 of Safety Performance Functions.

113 Polus and Mattar-Habib (2004) proposed the first global consistency model, which was
114 based on the hypothesis that the larger the speed variations along a road segment, the greater the
115 likelihood of crash occurrence. In this regard, the model was defined through the following
116 variables: relative area (R_a) and operating speed deviation (σ).

$$R_a = \frac{\sum_i^n a_i}{L} \text{ (m/s)} \quad (1)$$

$$\sigma = \sqrt{\frac{\sum_i^n (V_{85i} - \overline{V_{85}})^2}{n}} \text{ (km/h)} \quad (2)$$

117 where a_i is the area bounded by the operating speed profile and the average operating speed (m²/s);
 118 L is the length of the homogeneous road segment (m); $\overline{V_{85}}$ is the average operating speed (km/h);
 119 V_{85i} is the operating speed at station i (km/h); and n is the number of distance intervals (n is equal
 120 to L when i is considered meter by meter).

121 In this way, Polus and Mattar-Habib (2004) defined a new consistency parameter (C_P),
 122 which qualifies the consistency level of a homogenous road segment as good, fair, or poor (Table
 123 1).

124 The same variables were also studied by Garach et al. (2014) on Spanish two-lane rural
 125 roads. As a conclusion, Garach et al. (2014) found that the consistency parameter proposed by
 126 Polus and Mattar-Habib (2004) was too conservative, so a new global consistency model was
 127 suggested (C_G) with the same consistency thresholds (Table 1).

128 Camacho-Torregrosa (2015) also calibrated a global consistency model in Spain relying on
 129 two operational parameters: average operating speed ($\overline{V_{85}}$) and average deceleration rate ($\overline{d_{85}}$).
 130 According to the proposed consistency parameter (C_C), an increase in consistency level usually
 131 leads to a greater $\overline{V_{85}}$ and/or a lower $\overline{d_{85}}$ (Table 1).

132 Recently, Llopis-Castelló et al. (2018b) proposed a new global consistency model based on
 133 the difference between the inertial operating speed profile (V_i) and the operating speed profile
 134 (V_{85}). The inertial operating speed represents drivers' expectancies and was defined at every station
 135 k of the road segment as the weighted average operating speed of the preceding 15 seconds,
 136 considering a linear weighting distribution. This weighting factor (w_j) ranges between 0 and 1,
 137 increasing linearly as station j gets closer to the critical section k . This speed calculation is carried

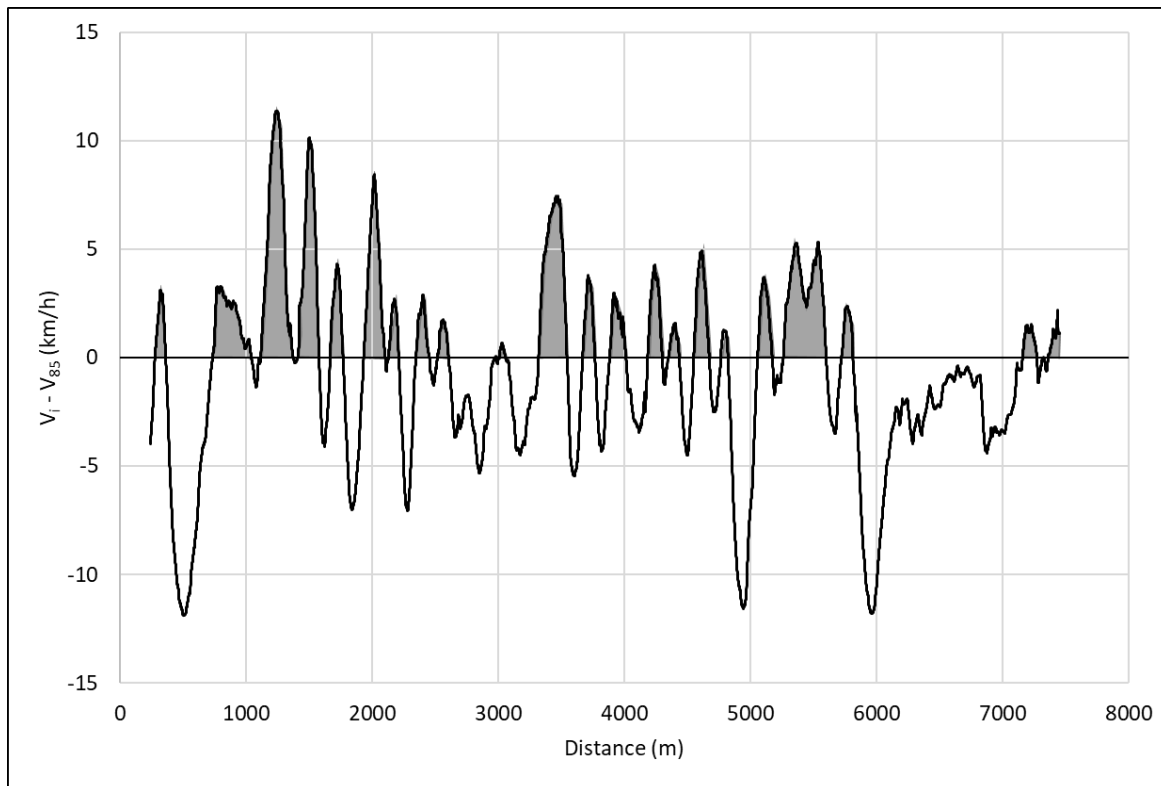
138 out considering intervals of 0.1 seconds through the following equation:

$$V_{i,k} = \frac{\sum w_j \cdot V_{85,j}}{\sum w_j} \quad (3)$$

139 where $V_{i,k}$ is the inertial operating speed (km/h) at station k ; $V_{85,j}$ is the operating speed at station
140 j ; and w_j is the weighting factor at point j .

141 Table 1 shows the consistency parameter defined by Llopis-Castelló et al. (2018b), which
142 is calculated from the positive differences between V_i and V_{85} considering the following variables
143 (Figure 1):

- 144 • A_+ : Area bounded by the positive differences between V_i and V_{85} .
- 145 • L_+ : Road segment length where the difference between V_i and V_{85} is positive.
- 146 • σ_+ : Standard deviation of the positive difference between V_i and V_{85} .



147
148 **FIGURE 1** Difference between V_i and V_{85} .

149 A low value of this consistency parameter is associated to lower differences between
150 drivers' expectancies and road behavior, leading to a higher consistency level.

151 **TABLE 1 Global consistency models**

Model	Consistency parameter (C)	Consistency level		
		Good	Fair	Poor
Polus and Mattar-Habib (2004)	$C_p = 2.808 \cdot e^{-0.278 \cdot R a \frac{\sigma}{3.6}}$	$C_p > 2$	$1 < C_p \leq 2$	$C_p \leq 1$
Garach et al. (2014)	$C_G = \frac{195.073}{\left(\frac{\sigma}{3.6} - 5.7933\right) \cdot (4.1712 - R_a) - 26.6047 + 6.7826}$	$C_G > 2$	$1 < C_G \leq 2$	$C_G \leq 1$
Camacho- Torregrosa (2015)	$C_C = \sqrt[3]{\frac{V_{85}}{d_{85}}} \text{ (s}^{1/3}\text{)}$	$C_C \geq 3.25$	$2.55 \leq C_C < 3.25$	$C_C < 2.55$
Llopis- Castelló et al. (2018b)	$C_L = \sqrt{\frac{A(+)}{L(+)} \cdot \sigma(+)} \text{ (km/h)}$	$C_L \leq 2.75$	$2.75 < C_L \leq 4.5$	$C_L > 4.5$

152

153 The current way of selecting a road design among a set of alternatives considers several items to
 154 maximize, such as cost, environment, road safety, functionality, etc. A concomitant maximization
 155 of all criteria is impossible in most occasions, so some techniques to balance between all factors
 156 are needed. Maximizing the benefit for one criterion normally impacts negatively in other criteria,
 157 so weighting factors have to be considered in order to provide the globally best solution.

158 1.3. Objective and hypothesis

159 All previous consistency models reveal that a higher speed variation along a road segment results
 160 in a lower consistency level. Likewise, winding road segments are prone to have a higher speed
 161 variation and, consequently, produce larger vehicle CO₂ emissions (Llopis-Castelló et al., 2018a).
 162 Merging up both statements, we can presume that lower CO₂ emissions might be expected on
 163 consistent roads under similar conditions of driver behavior, vehicle, and traffic, which are other
 164 important factors affecting this phenomenon.

165 Since minimizing CCR (and hence, enhancing the road design) and minimizing emissions
 166 have proven to be concomitant effects, the exploration of the relationship between consistency and
 167 road emissions due to horizontal design are suggested. A good level of correlation would ease the
 168 maximization of the alternative selection in the predesign process, since two important goals would

169 be achieved in the same direction.

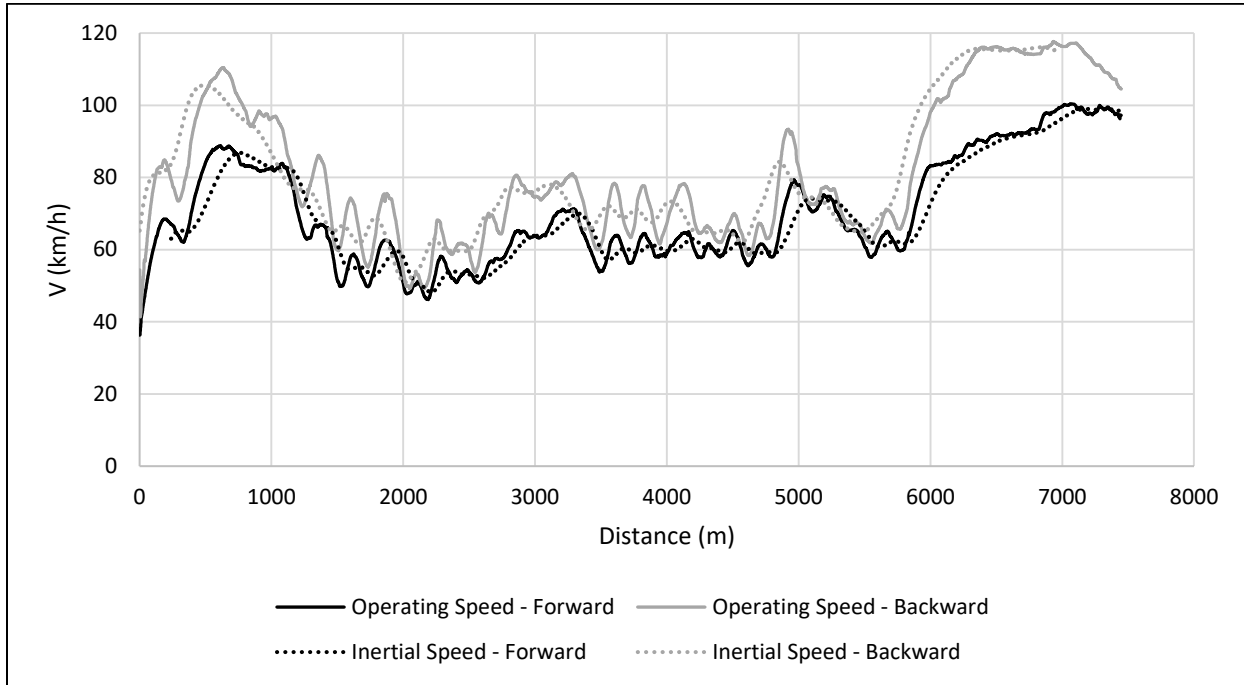
170 Therefore, the main objective of this research is to examine the influence of geometric
171 design consistency on vehicle CO₂ emissions through actual continuous speed profiles collected
172 on 15 two-lane rural road sections. The study focuses on CO₂ emissions because this Greenhouse
173 Gas is the main responsible for global warming and, in addition, this is directly proportional to
174 vehicle fuel consumption.

175 **2. METHODOLOGY**

176 This study was developed by analyzing the relationship between geometric design consistency and
177 CO₂ vehicle emissions on 15 two-lane rural road sections localized in the Valencian Region
178 (Spain). Speed data and estimated vehicle CO₂ emissions used in this research are part of a
179 previous study developed by the same research group (Llopis-Castelló et al., 2018a).

180 These speed profiles correspond to actual drivers and were gathered during a naturalistic
181 data collection following the methodology proposed by Pérez-Zuriaga et al. (2010). Two
182 checkpoints were localized at the beginning and at the end of each road section. Then, every
183 vehicle travelling along the road was stopped, and drivers were asked to participate in the speed
184 data collection. If the driver accepted, a 1 Hz pocket-sized GPS was placed on the vehicle. As a
185 result of the data collection, the individual continuous speed profile for each driver was obtained.

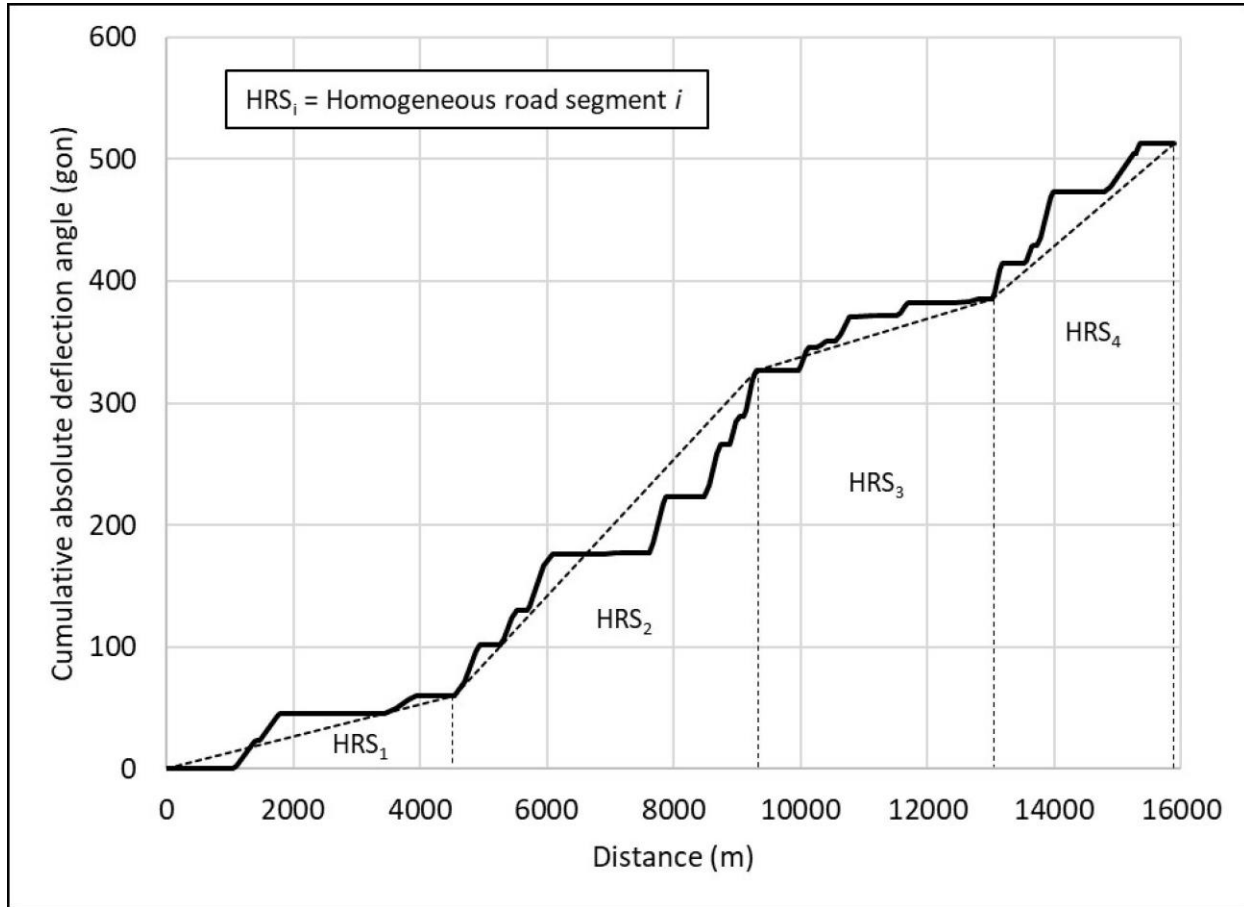
186 Then, free-flow conditions were checked by means of the procedure proposed by Pérez-
187 Zuriaga et al. (2013), which relies on the hypothesis that every single driver behaves according to
188 a specific speed percentile. Thus, non-free flow road sections are associated to sudden variations
189 in its usual operating percentile. After removing individual non-free-flow sections, the operating
190 speed profile of each road section was estimated for both forward and backward direction (Figure
191 2).



192
193 **FIGURE 2 Speed profiles for both forward and backward direction.**

194 Additionally, the studied road sections were divided into homogeneous road segments in
195 order to assess geometric design consistency and estimate CO₂ emissions. In this regard, a
196 homogeneous road segment is one which has a constant traffic volume, does not contain major
197 intersections and interchanges that might significantly influence drivers' behavior, and presents a
198 similar geometric behavior regarding the German method, which is based on the parameter
199 Curvature Change Rate (*CCR*).

200 The German method aims at identifying homogeneous road segments by depicting the
201 cumulative absolute deflection angle versus the road station, so that the parameter *CCR* is
202 represented by the slope (Figure 3). In this way, a homogenous road segment is one which has an
203 approximately constant value for *CCR*, i.e., a similar slope.



204
205 **FIGURE 3 Homogeneous road segment identification based on the German method.**

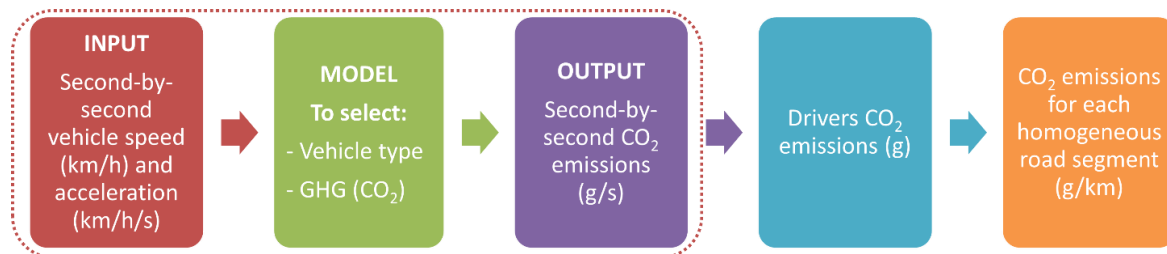
206 As a result, 47 homogeneous road segments were identified. These have a length ranging
207 from 955 m to 7,864 m, with an average length of 2,183 m. Their Curvature Change Rates (*CCR*)
208 range from 0 gon/km to 645 gon/km with an average value of 156 gon/km. Regarding cross-
209 section, lane width ranges between 3.00 and 3.50 m, whereas shoulder width varies between 0.5
210 and 1.50 m. Their longitudinal grade is not greater than 5%.

211 Finally, vehicle CO₂ emissions and geometric design consistency were estimated and the
212 influence of the geometric design consistency on vehicle emissions was studied by comparing the
213 level of consistency of each homogeneous road segment with its CO₂ emission rate (g/km).

214

215 3. ESTIMATION OF VEHICLE CO₂ EMISSIONS AND CONSISTENCY LEVEL

216 Vehicle CO₂ emissions were estimated previously by Llopis-Castelló et al. (2018a) using the
 217 microscopic model VT-Micro (Ahn et al., 2002), which estimates GHG emission rates for different
 218 categories of vehicle considering second-by-second vehicle speed (km/h) and acceleration
 219 (km/h/s) as input variables (Figure 4). This model was preferred over other softwares such as
 220 MOBILE5a, MOBILE6, and CMEM because VT-Micro is able to provide more reliable
 221 estimations and is frequently updated (Rakha et al., 2003).



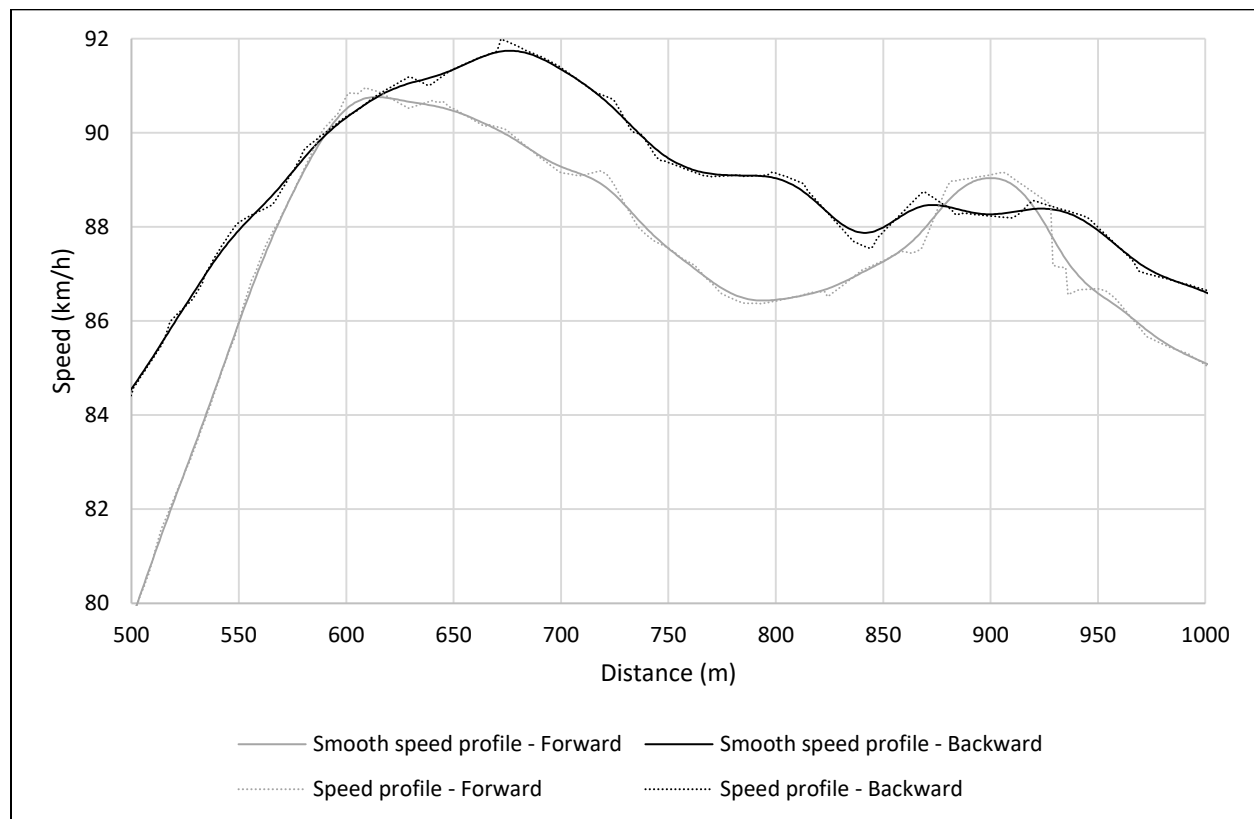
222
 223 **FIGURE 4 VT-Micro model procedure.**

224 The vehicles which took part in the field data collection were largely sedan-type vehicles
 225 and van-type vehicles which correspond to the following categories of vehicles according to Rakha
 226 et al. (2004): LDV3, LDV4, and LDV5. Therefore, CO₂ emissions were only calculated for these
 227 vehicle categories.

228 Finally, the CO₂ emission rate (g/km) for each homogeneous road segment was estimated
 229 as the average of all drivers CO₂ emissions divided by the length of the homogeneous road
 230 segment, taking into account both forward and backward directions (Table 2).

231 On the other hand, the consistency level for each homogeneous road segment was obtained
 232 considering the global consistency models developed by Polus and Mattar-Habib (2004), Garach
 233 et al. (2014), Camacho-Torregrosa (2015), and Llopis-Castelló et al. (2018b). All their consistency
 234 parameters are based on different consistency variables extracted from the operating speed profile
 235 (Table 1).

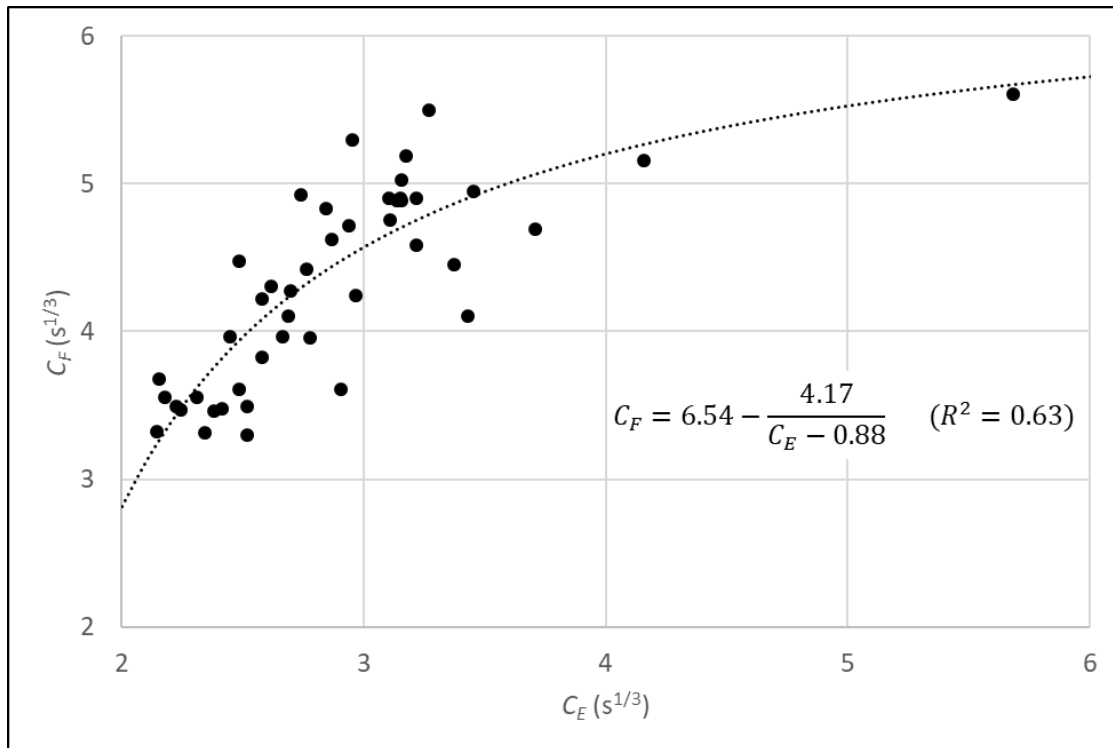
236 However, the model developed by Camacho-Torregrosa (2015) was developed only for
 237 operating speed profiles calculated with speed models (estimated speed profile). Field-extracted
 238 operating speed profiles (actual speed profile) present much more noise than estimated ones, so
 239 results might be biased. Thus, a smoothing algorithm was applied to field operating speed profiles
 240 prior to this consistency model to ensure a feasible average deceleration rate (Figure 5). This
 241 algorithm estimated the speed at each station as a function of a moving average speed.



242
 243 **FIGURE 5 Smooth speed profiles.**

244 Additionally, the consistency thresholds proposed by Camacho-Torregrosa (2015) were
 245 redefined for the same reason. To do this, the consistency parameters obtained through the actual
 246 speed profiles collected in field (C_F) were compared with those obtained considering estimated
 247 speed profiles (C_E), which were calculated by means of the speed models proposed by Pérez-
 248 Zuriaga (2012).

249 Figure 6 shows the close relationship between C_F and C_E . As a result, an expression that
 250 relates C_F to C_E were calibrated and new consistency thresholds were defined taking into account
 251 the previous ones shown in Table 1. To this regard, a homogeneous road segment has a good
 252 consistency level when C_F is greater than $4.8 \text{ s}^{1/3}$, a poor consistency level when C_F is lower than
 253 $4 \text{ s}^{1/3}$, and a fair consistency level otherwise.



254
 255 **FIGURE 6 Analysis of the consistency thresholds proposed by Camacho-Torregrosa (2015).**

256 Table 2 contains the consistency parameter and the qualitative consistency level for each
 257 homogeneous road segment according to all consistency models.

258 **TABLE 2 CO₂ emission rates and consistency**

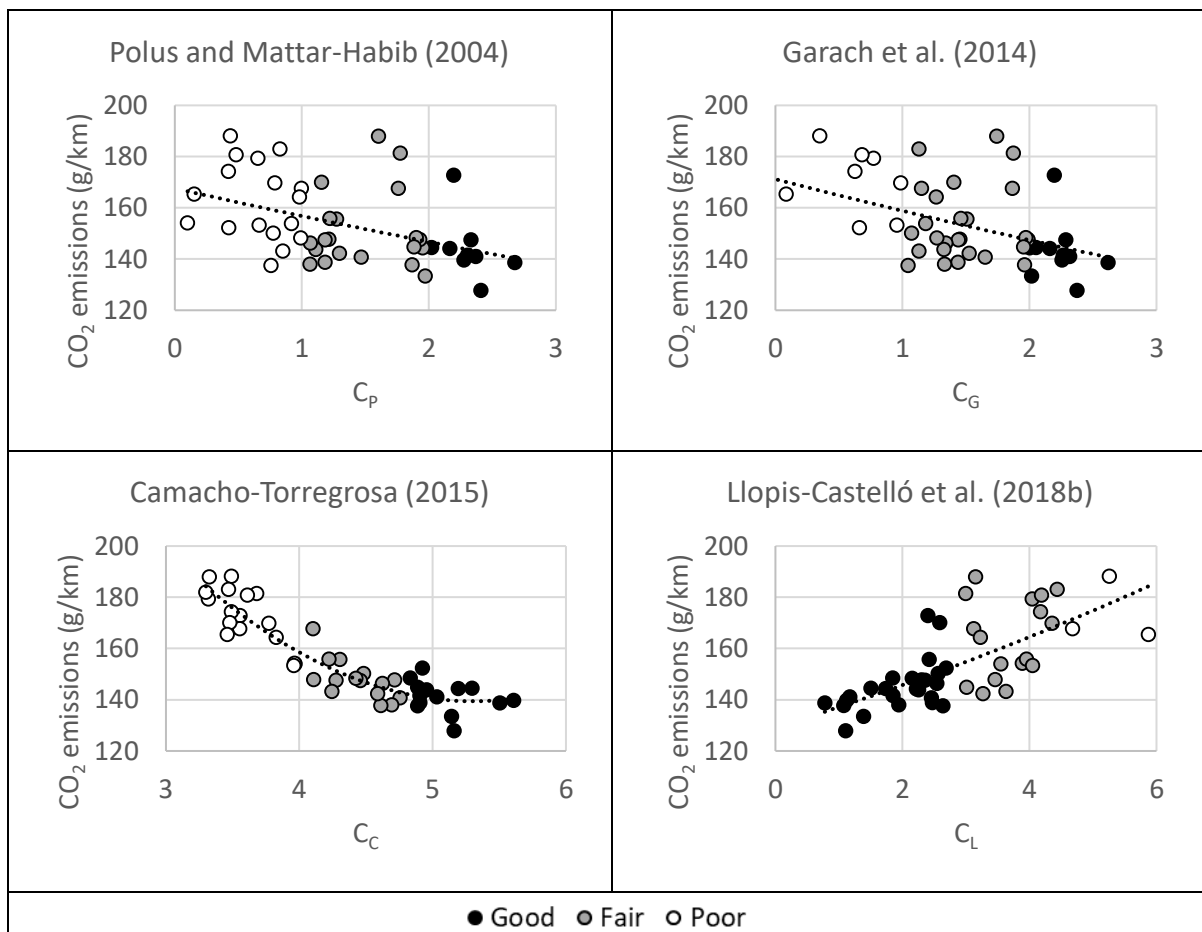
Homogeneous Road Segment	Length (m)	CCR (gon/km)	CO ₂ (g/km)			Polus and Mattar-Habib (2004)		Garach et al. (2014)		Camacho-Torregrosa (2015)		Llopis-Castelló et al. (2018b)	
			LDV3	LDV4	LDV5								
1.1	1477	68	165.40	150.50	202.45	0.1532	Poor	0.0857	Poor	3.4598	Poor	5.8718	Poor
1.2	1292	645	187.96	169.52	237.19	1.6040	Fair	1.7425	Fair	3.3241	Poor	3.1513	Fair
1.3	1093	292	167.65	152.34	208.39	1.7595	Fair	1.8661	Fair	3.5533	Poor	3.1243	Fair
1.4	955	628	172.83	156.52	216.54	2.1948	Good	2.1966	Good	3.5551	Poor	2.4016	Good
1.5	1059	245	170.07	154.16	211.04	1.1558	Fair	1.4059	Fair	3.4784	Poor	2.5865	Good

1.6	1519	36	147.50	137.65	175.00	1.1828	Fair	1.4396	Fair	4.4567	Fair	2.3016	Good
2.1	1592	4	167.68	156.32	194.86	0.9955	Poor	1.1508	Fair	4.1015	Fair	4.6799	Poor
2.2	2392	534	179.36	162.35	222.14	0.6535	Poor	0.7714	Poor	3.3163	Poor	4.0450	Fair
2.3	2583	347	188.16	168.08	234.72	0.4366	Poor	0.3500	Poor	3.4902	Poor	5.2579	Poor
2.5	2069	254	180.80	164.20	222.34	0.4819	Poor	0.6829	Poor	3.6112	Poor	4.1897	Fair
2.6	5937	103	169.82	155.50	205.70	0.7882	Poor	0.9860	Poor	3.7709	Poor	4.3571	Fair
3.1	2691	20	137.99	129.05	161.86	1.0650	Fair	1.3324	Fair	4.6921	Fair	1.9407	Good
3.2	2689	136	147.85	136.90	177.07	1.2119	Fair	1.4529	Fair	4.1064	Fair	3.4578	Fair
3.3	1721	549	174.33	157.62	216.03	0.4219	Poor	0.6272	Poor	3.4891	Poor	4.1733	Fair
3.4	2867	151	153.90	141.91	184.97	0.9171	Poor	1.1839	Fair	3.9672	Poor	3.5499	Fair
3.5	2396	0	137.55	129.08	160.53	0.7563	Poor	1.0435	Fair	4.8871	Good	2.6363	Good
4.1	2321	317	150.16	137.97	183.56	0.7745	Poor	1.0736	Fair	4.4804	Fair	2.5619	Good
4.2	1625	139	148.42	137.84	178.32	1.9006	Fair	1.9717	Fair	4.8313	Good	1.8456	Good
5.1	1374	0	137.75	129.86	159.28	1.8687	Fair	1.9607	Fair	4.6121	Fair	1.0766	Good
5.2	1879	50	127.84	119.40	147.28	2.4101	Good	2.3754	Good	5.1607	Good	1.1073	Good
5.3	7864	5	133.48	125.66	154.50	1.9705	Fair	2.0171	Good	5.1407	Good	1.3883	Good
6.1	1082	220	154.14	141.77	186.56	0.1002	Poor	-0.1070	Poor	3.9641	Poor	3.8874	Fair
6.2	2968	60	143.15	132.76	169.56	0.8513	Poor	1.1296	Fair	4.2431	Fair	3.6301	Fair
6.3	3627	173	153.30	141.39	184.36	0.6627	Poor	0.9558	Poor	3.9577	Poor	4.0517	Fair
7.1	1604	0	139.64	131.38	162.94	2.2773	Good	2.2572	Good	5.6071	Good	1.1216	Good
7.2	2833	44	143.83	134.44	169.43	1.1086	Fair	1.3277	Fair	4.9530	Good	2.2575	Good
8.1	1739	81	141.07	132.54	165.84	2.3698	Good	2.3168	Good	5.0305	Good	1.1747	Good
8.2	1138	14	138.70	131.84	159.87	2.6772	Good	2.6197	Good	5.5033	Good	0.7770	Good
8.3	2349	73	142.35	133.59	167.64	1.2973	Fair	1.5264	Fair	4.5874	Fair	3.2706	Fair
9.1	1264	156	164.34	149.68	203.86	0.9814	Poor	1.2663	Fair	3.8255	Poor	3.2273	Fair
9.2	1663	596	181.41	164.12	228.85	1.7757	Fair	1.8727	Fair	3.6786	Poor	2.9986	Fair
9.3	2226	406	183.08	165.45	229.52	0.8282	Poor	1.1302	Fair	3.4679	Poor	4.4371	Fair
9.4	1006	125	181.87	164.25	225.94	0.0933	Poor	-1.1499	Poor	3.2991	Poor	5.8896	Poor
10.1	1775	83	155.87	143.33	189.09	1.2207	Fair	1.4608	Fair	4.2230	Fair	3.9538	Fair
10.2	2141	31	146.27	135.53	175.02	1.0671	Fair	1.3420	Fair	4.6234	Fair	2.5445	Good
11.1	2529	157	144.49	134.11	174.02	2.0222	Good	2.0518	Good	5.2943	Good	1.5044	Good
11.2	1239	9	144.79	134.78	173.79	1.8841	Fair	1.9538	Fair	4.8881	Good	3.0104	Fair
12.1	4638	21	138.77	129.15	164.17	1.1818	Fair	1.4376	Fair	4.9045	Good	2.4739	Good
12.2	2843	59	141.58	131.99	168.79	2.3036	Good	2.2660	Good	4.9041	Good	1.8530	Good
13.1	1054	42	152.34	140.92	183.74	0.4255	Poor	0.6614	Poor	4.9226	Good	2.6877	Good
13.2	1529	80	148.32	137.17	179.21	0.9904	Poor	1.2713	Fair	4.4242	Fair	2.1560	Good
13.3	1915	9	144.14	134.27	171.99	2.1650	Good	2.1594	Good	4.9001	Good	2.2186	Good
13.4	1128	62	147.53	135.95	179.93	2.3307	Good	2.2856	Good	4.2750	Fair	2.3563	Good
14.1	1832	15	144.38	135.64	169.76	1.9500	Fair	2.0031	Good	5.1923	Good	1.7371	Good
14.2	1702	172	155.70	143.37	188.79	1.2732	Fair	1.5076	Fair	4.3042	Fair	2.4194	Good
14.3	2457	92	147.73	137.05	177.72	1.9296	Fair	1.9885	Fair	4.7141	Fair	2.2999	Good
15.1	2930	36	140.76	130.82	169.08	1.4683	Fair	1.6529	Fair	4.7541	Fair	2.4592	Good

259 **4. ANALYSIS AND RESULTS**

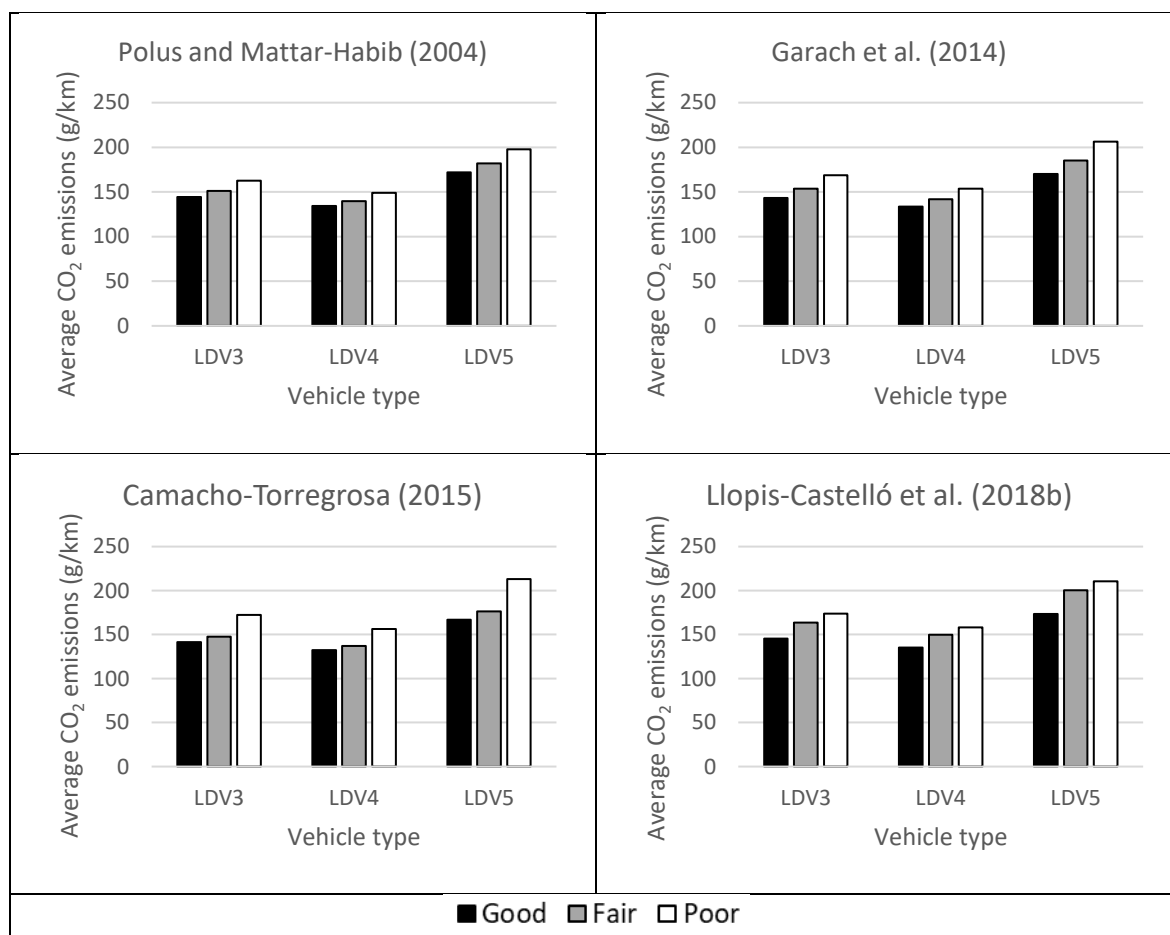
260 The analysis of this study was developed through the comparison between geometric design
 261 consistency and vehicle CO₂ emissions considering the above defined global consistency models.

262 The relationship between each consistency parameter and CO₂ emissions for each
 263 homogeneous road segment and vehicle type (LDV3, LDV4, and LDV5) was studied with a
 264 descriptive analysis. Figure 7 shows this relationship for LDV3 vehicle type. Although the global
 265 consistency models proposed by Polus and Mattar-Habib (2004) and Garach et al. (2014) showed
 266 a greater variability than those models defined by Camacho-Torregrosa (2015) and Llopis-Castelló
 267 et al. (2018b), all of them result in higher vehicle CO₂ emissions as the consistency level gets
 268 lower. The same conclusion was observed for all vehicle types (LDV4 and LDV5).



269 **FIGURE 7 Geometric design consistency Vs. Vehicle CO₂ emissions (LDV3 vehicle type).**

270 Additionally, the average CO₂ emissions was calculated for any consistency level
 271 considering each global consistency model and vehicle type (Figure 8). Although the level of CO₂
 272 emissions is different for each vehicle type, all of them show a similar trend. A consistent road
 273 segment leads to lower vehicle CO₂ emissions, supporting the above-mentioned results.



274 **FIGURE 8 Consistency level Vs. Average CO₂ emissions.**

275 Finally, a hypothesis test was performed to identify whether the average CO₂ emissions for
 276 the different consistency levels are statistically different or not (Table 3). For this, the following
 277 hypotheses were formulated: (a) Null hypothesis (H₀): the average CO₂ emissions for all
 278 consistency levels may be equal; (b) Alternative hypothesis (H₁): the average CO₂ emissions for
 279 the different consistency levels differ from each other. The confidence level considered was 95%.

280 The P-value for all global consistency models and vehicle types was lower than 0.05.

281 Therefore, we can conclude – at a 95% confidence level – that the average CO₂ emissions differ
 282 with the consistency level, regardless the consistency criterion. In addition, the Least Significant
 283 Difference (LSD) intervals were analyzed. Most of these intervals did not overlap, strengthening
 284 the previous statement.

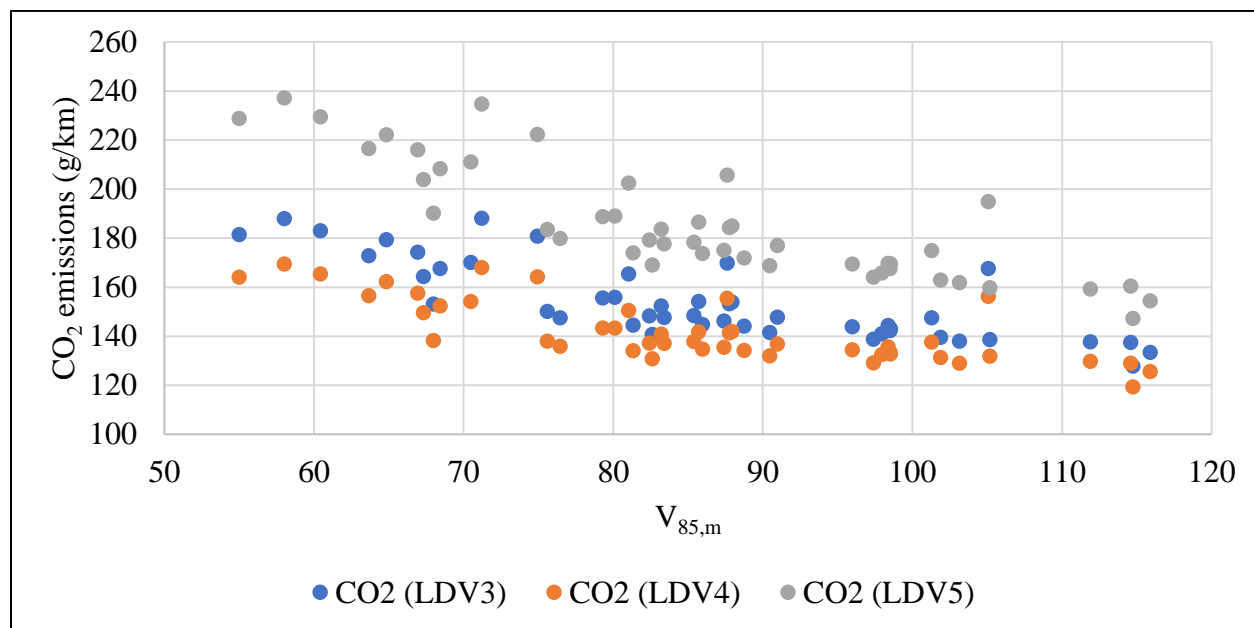
285 **TABLE 3 Statistical analysis: Average CO₂ emissions Vs. Consistency level.**

Consistency model	Vehicle type	Statistical parameter	Consistency		
			Good	Fair	Poor
Polus and Mattar-Habib (2004)	LDV3	Average (g/km)	144.202	151.027	163.761
		Std. dev. (g/km)	12.074	14.787	15.2649
		F	6.49		
		P-value	0.0034		
	LDV4	Average (g/km)	134.222	139.741	149.83
		Std. dev. (g/km)	9.62865	11.752	12.3303
		F	6.42		
		P-value	0.0036		
	LDV5	Average (g/km)	171.911	181.801	199.447
Std. dev. (g/km)		19.1685	22.6816	21.9267	
F		5.66			
P-value		0.0065			
Garach et al. (2014)	LDV3	Average (g/km)	143.243	153.495	169.953
		Std. dev. (g/km)	11.2747	14.6771	13.1526
		F	10.2		
		P-value	0.0002		
	LDV4	Average (g/km)	133.572	141.687	154.658
		Std. dev. (g/km)	9.01322	11.8692	10.4346
		F	9.82		
		P-value	0.0003		
	LDV5	Average (g/km)	170.132	185.301	208.398
Std. dev. (g/km)		17.9229	22.0204	18.7019	
F		9.26			
P-value		0.0004			
Camacho-Torregrosa (2015)	LDV3	Average (g/km)	141.402	147.774	172.26
		Std. dev. (g/km)	5.92498	7.69506	11.4213
		F	55.61		
		P-value	0.000		
	LDV4	Average (g/km)	132.202	137.157	156.433
		Std. dev. (g/km)	5.11312	6.76259	9.22302
		F	49.43		
		P-value	0.000		
	LDV5	Average (g/km)	166.998	176.512	212.977
Std. dev. (g/km)		9.29243	10.0512	16.8076	
F		58.88			
P-value		0.000			
Llopis-Castelló et al (2018b)	LDV3	Average (g/km)	173.393	200.416	214.493
		Std. dev. (g/km)	15.3528	22.9207	18.8907
		F	18.27		
		P-value	0.000		
	LDV4	Average (g/km)	135.13	149.837	159.789
		Std. dev. (g/km)	7.71623	12.4701	7.89258
		F	18.61		
		P-value	0.000		
	LDV5	Average (g/km)	173.393	200.416	214.493
Std. dev. (g/km)		15.3528	22.9207	18.8907	
F		15.59			
P-value		0.000			

286 5. DISCUSSION

287 This research highlights that geometric design consistency influences vehicle CO₂ emissions. In
 288 this regard, lower emissions are expected on consistent road designs.

289 The consistency models developed by Camacho-Torregrosa (2015) and Llopis-Castelló et
 290 al. (2018b) showed a greater correlation with CO₂ emissions than those proposed by Polus and
 291 Mattar-Habib (2004) and Garach et al. (2014). This higher correlation could be explained due to
 292 the more direct connection to speed and its standard deviation (σ) that these models have.
 293 According to Llopis-Castelló et al. (2018a), the average mean speed and σ were operational
 294 parameters highly connected to emissions. Figure 9 shows the relationship between CO₂ emissions
 295 and the average operating speed. It seems evident that a road segment which presents a higher
 296 operating speed would require less time for drivers to perform along it. Therefore, the final
 297 emission rate is lower for faster road segments.

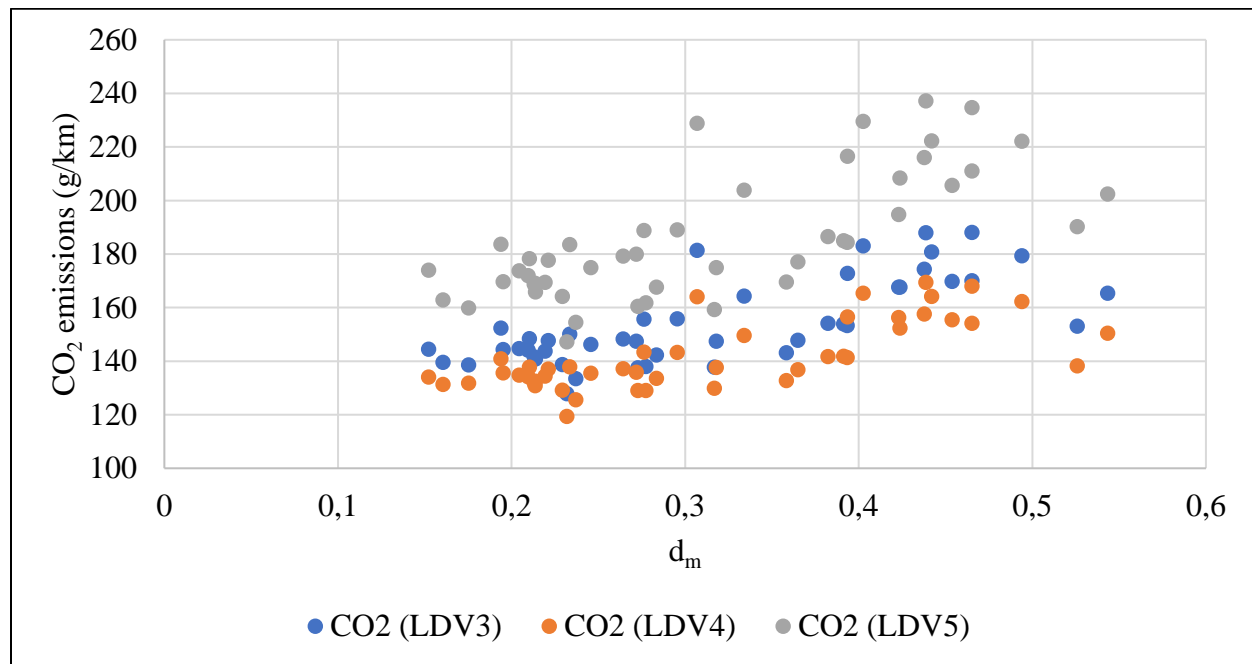


298

299 **FIGURE 9 CO₂ emissions as a function of the average operating speed.**

300 However, speed is not the only factor causing emissions to rise. Operating speed variability is also
 301 a major contributor. Figure 10 shows how speed variation (expressed in terms of average

302 deceleration rate) impacts emissions. As expected, higher decelerations are linked to higher
 303 emissions. While decelerations per se are not the main contributor (should be accelerations), this
 304 parameter has been examined since it is one of the components of Camacho-Torregrosa's
 305 consistency model. Anyway, higher deceleration rates are generally linked to medium to long
 306 tangents followed by sharp curves, so high acceleration rates are also expected to be found.



307

308 **FIGURE 10 CO₂ emissions as a function of the average deceleration rate.**

309 Both Camacho-Torregrosa (2015) and Llopis-Castelló et al. (2018b) consistency models are very
 310 well connected to these operational parameters, so increasing the consistency level on a road
 311 segment would presumably result in higher average operating speeds and/or lower speed
 312 variability. Since maximizing consistency increases the safety level of the road segment, its
 313 maximization would also work towards minimizing emissions.

314 Although both consistency models can be used for that purpose, Camacho-Torregrosa's
 315 model seems to depict a little bit better the relationship for poor-consistent road segments (lower
 316 noise), while Llopis-Castelló et al.'s one works better for medium-to-good consistent road

317 segments (see Figure 7, bottom).

318 An analytic quantification of the amount of emissions that can be attributed to the
319 horizontal alignment cannot be performed, since there are no boundaries for good and poor
320 consistency levels. However, we can compare the emission rates among consistency levels.
321 According to Camacho-Torregrosa's consistency model, a poor consistency road segment presents
322 an emission rate 27% higher than for a good consistency level (LDV5). For Llopis-Castelló's
323 model, this difference is 21%.

324 It is also necessary to highlight that all these conclusions have been obtained using
325 operating speed profiles, i.e., free-flow speed distributions. Since a 5-second interval is needed to
326 ensure free-flow conditions, the validity of these conclusions extends up to 720 pc/h/direction. In
327 fact, the maximum volume is even lower, since it has been determined following an ideal,
328 completely uniform distribution of vehicles. However, this hourly volume demand is considered
329 to be high enough to cover most cases in Spain – especially the poor-consistent ones.

330 The authors are working on the expansion of these conclusions to road segments showing
331 a higher demand. This will be done by means of combining free-flow distributions for other
332 percentiles (García-Jiménez et al., 2016) with normal-distributed time intervals, car-following, and
333 passing models in a microsimulation environment.

334 **6. CONCLUSIONS**

335 There are different factors involved in GHG emissions production derived from the transportation
336 sector, such as driver behavior, vehicle, traffic operation, and road design. Although the policy
337 measures to mitigate emissions are increasing, none of them consider the road geometric design
338 as a potential measure to reduce air pollution.

339 Thus, the objective of this research was to analyze the relationship between geometric

340 design consistency and vehicle CO₂ emissions using actual continuous speed data collected on 47
341 homogeneous road segments.

342 Although the CO₂ emission level was different for each vehicle type (LDV3, LDV4, and
343 LDV5), all of them showed an increase in vehicle CO₂ emissions as the consistency level of a
344 homogeneous road segment decreases. To this regard, statistically significant differences were
345 identified among the CO₂ emission levels associated to the different consistency levels for all
346 vehicle types. Despite absolute values are different as a function of the vehicle type, these
347 differences are homothetical, according to the model provided by Ahn et al. (2002). Thus, a
348 consistent road leads to lower vehicle CO₂ emissions, regardless the vehicle type.

349 Although the research has been carried out for free-flow conditions, it is worth to say that
350 these are the prevailing conditions for Spanish two-lane rural roads. Higher traffic volumes can be
351 found at urban or suburban environments, which are out of scope of this study.

352 As vehicle CO₂ emissions are directly proportional to vehicle fuel consumption, the
353 outcomes obtained are also applicable to vehicle fuel consumption. Therefore, the design of
354 consistent roads allows, in addition to maximize road safety, to help to achieve more
355 environmentally sustainable highways, reducing CO₂ emission production.

356 Since these models only consider the horizontal alignment, further research is suggested to
357 analyze the influence of the vertical alignment and the interaction between vehicles on GHG
358 emissions. In addition, a wider perspective of the road lifecycle assessment could be further
359 explored by including the construction stage in the analysis. While a consistent road segment
360 derives into lower emissions at an operational level; the construction stage would normally involve
361 an increase in earthworks – especially for mountainous highways – so the total emissions for the
362 whole lifecycle might be even higher, compared to a low-consistent alternative.

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