

Highlights

Impact of Mileage on Particle Number Emission Factors for EURO5 and EURO6 Diesel Passenger Cars

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- Particle number measurements on 757 diesel passenger cars from the field, homologated according to the EURO5 and EURO6 emission standards.
- Fleet average emission factors are strongly increased due to the presence of high emitters.
- The Handbook Emission Factors for Road Transport strongly underestimate the particle number emission factors of the tested fleet.
- Mileage has a significant impact on the particle number emissions.

Impact of Mileage on Particle Number Emission Factors for EURO5 and EURO6 Diesel Passenger Cars

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ABSTRACT

Air quality is a growing concern worldwide because of its impacts on both the environment and the human health. The road transport sector is a major contributor to this poor air quality. To reduce the emission of particulate matter, all diesel passenger cars were equipped with diesel particulate filters since the EURO5b emission standard. Unfortunately, these filters can be damaged or intentionally removed during the lifetime of a vehicle. This work presents the particle number emission factors for EURO5 and EURO6 diesel passenger cars, based on the measurements of 757 vehicles. These measurements were performed at low idle, which shows a high correlation to particle number emission factors obtained during homologation cycles or real-driving emission measurements. The results show that the average Particle Number (PN) emission factors are highly impacted by high emitters present in the fleet and that the mileage has a significant impact on the PN emission factors. Finally, the estimated PN emission factors based on low idle measurements were higher by a factor 5.6 for EURO5a, 2.5 for EURO5b and 5.5 for EURO6, compared to their respective HBEFA ([Handbook Emission Factors for Road Transport](#)) emission factors.

1. Introduction

1 Air quality is a major environmental and health issue in
2 many places over the world. In Europe, air quality limit val-
3 ues have been defined for PM_{2.5} and PM₁₀ (particulate mat-
4 ter with a diameter smaller than 2.5 and 10 µm, respectively):
5 the PM_{2.5} yearly average must not exceed 25 µg/m³ while
6 PM₁₀ values must respect a yearly average of 40 µg/m³ and
7 a 24-hour average of 50 µg/m³ (this 24-hour average can be
8 exceeded 35 times per year). In 2016, the yearly PM_{2.5} limit
9 was not respected at 5% of the European reporting stations
10 while the PM₁₀ daily limit was exceeded at 19% of these
11 stations. Also in 2016, long-term exposure to PM_{2.5} caused
12 422 000 premature deaths in Europe [9].

13 Additionally to these EU limits, the World Health Or-
14 ganization (WHO) defined Air Quality Guidelines (AQG)
15 which are more strict than the current EU limits. These
16 guidelines recommend that PM_{2.5} remains below 10 µg/m³
17 for the annual average and below 25 µg/m³ for the 24-hour
18 average. Regarding PM₁₀, the guidelines are 20 µg/m³ for
19 the annual mean and 50 µg/m³ for the daily mean. These
20 more stringent yearly values were exceeded at 68% of the
21 European reporting stations for the PM_{2.5} and at 48% of the

stations for PM₁₀ [9]. These air quality issues are even more
severe in urban areas as between 74 and 85% of urban popu-
lations in the EU-28 are exposed to yearly PM_{2.5} concentra-
tions above the WHO AQG since 2014.

Regarding the ultrafine particles (particulate matter with
a diameter smaller than 0.1 µm), there is still a knowledge
gap regarding their impact on human health which explains
the absence of air quality guidelines for this type of pollutant
[13]. Ultrafine particles may be more **harmful** than PM_{2.5}
and PM₁₀ because their smaller sizes allow them to pene-
trate deeper inside the respiratory system and to translocate
to different organs [13].

The road transport sector is a major contributor to this
poor air quality as it accounted for 11% of PM_{2.5} and for
10% of PM₁₀ in 2016 for the EU-28 countries [9]. Regarding
the ultrafine particles, the road transport was responsible for
around 40% of the total emissions in Europe in 2010 [15].

Emission models, such as HBEFA ([Handbook Emission
Factors for Road Transport](#)) or COPERT ([COmputer Pro-
gramme to calculate Emissions from Road Transport](#)), are
used by public authorities to estimate the emissions of the
road transport sector. They use as inputs: environmental
data, fleet characteristics and activity data. Based on the model
parameters such as the Emission Factors (EFs) (amount of
pollutant emitted per travelled kilometer) and the degrada-
tion factors (to account for deterioration of emissions with
mileage), the model estimates the emissions of this specific

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49 fleet. These emission models are used to compute the emis- 106
50 sion inventories for a city, a region or a country. They are 107
51 also useful to assess the potential impact of mobility poli- 108
52 cies such as Low Emission Zones progressively banning the 109
53 most polluting vehicles. 110

54 The accuracy of these model outputs is largely affected 111
55 by uncertain model parameters and model inputs. The ma- 112
56 jor uncertainty source comes from the EFs used in this model 113
57 [18]. Initially these EFs were only based on laboratory tests 114
58 but now they also include real-world results such as remote 115
59 sensing or real driving testing using a Portable Emissions 116
60 Measurement System (PEMS). Real driving emission mea- 117
61 surements provide very accurate results for a limited number 118
62 of vehicles but are very expensive and time consuming. 119

63 A major drawback of these emission models is that they 120
64 use averages for all the inputs and model parameters. Be- 121
65 cause their average value is considered, they fail to a great 122
66 extend to represent the very large variability of local pollu- 123
67 tants because these pollutant emissions are extremely non- 124
68 linear. Several studies also show that the real-world emis- 125
69 sions are non-linearly spread from below homologation val- 126
70 ues up to more than a factor 10 above [3, 6]. Instead of using 127
71 averages, these models could use probability distributions
72 for their inputs and model parameters to better represent their
73 wide spread and take into account the non-linearities of Part-
74 icle Number (PN) emissions. The usage of averages is usu-
75 ally motivated by the lower computational cost which is re-
76 quired compared to using distributions. Nevertheless, the
77 curse of dimensionality generally limit the application of un-
78 certainty in complex systems. Recent techniques take ad-
79 vantage of the sparsity of such systems to significantly re-
80 duce the computational cost (up to 10 times) of uncertainty
81 propagation [1]. Including uncertainties, in particular for
82 highly non-linear pollutants, would improve the predictions
83 and better guide policy makers.

84 Previous studies [17, 4] proved that actual fleet emis- 130
85 sions can be highly affected by high emitters. In the case 131
86 of particulate matter for diesel vehicles, Diesel Particulate 132
87 Filter (DPF) became standard since the EURO5a emission 133
88 standard in 2009 and became mandatory since the EURO5b 134
89 emission standard in 2011, in order to lower their emissions. 135
90 Unfortunately, high emitters among those recent diesel ve- 136
91 hicles can be explained by damaged or voluntarily removed 137
92 DPFs. Because current Periodic Technical Inspection (PTI) 138
93 cannot detect such high emitters of particulate matter, their 139
94 DPFs are not repaired or replaced and the contribution of 140
95 these high emitters could increase the average fleet emissions 141
96 by a factor 30 [4]. 142

97 In this context, recent studies were conducted in Bel- 143
98 gium, Germany, the Netherlands and Switzerland to assess 144
99 the possibility of implementing a new test procedure to de- 145
100 tect these high emitters of particulate matter emissions. We 146
101 developed a test procedure which consists of measuring the 147
102 PN from a diameter of 23 nm and above, such as for the ho- 148
103 mologation but instead of following a cycle as during the ho- 149
104 mologation, the test would be performed at low idle. While 150
105 not replacing costly and time consuming homologation tests,

low idle PN measurements show high correlation with their 106
results [8]. This new procedure will be implemented in Bel- 107
gium as of 2021 while Germany and Netherlands should im- 108
plement a similar procedure in 2021. 109

110 This paper first explains the measurement procedure, the 111
112 data collected during its development (i.e. particulate matter 113
114 concentration of 757 EURO5 and EURO6 diesel passenger 115
116 cars during a low idle test) and the measurement devices. 117
118 Then, the correlation between PN concentrations measured 119
120 at low idle and PN emission factors is discussed. Finally, the 121
122 computed PN emission factors are provided for the tested ve- 123
124 hicles from the actual fleet together with an analysis of influ- 125
126 encing factors such as the emission standard and the mileage 127
of the vehicles. Using this large database, this paper extends
the current use of PN emission factor averages to distribu-
tions to be used in emission models, taking into account the
small fraction of high emitting vehicles having a major im-
pact on the fleet average. Ultimately, the goal of this paper is
to provide a better characterization of the emissions of these
EURO5 and EURO6 diesel vehicles since they are becom-
ing the only diesel vehicles allowed in many Low Emission
Zones implemented across the world.

2. Methodology 128

2.1. Test procedure and measurement devices 129

130 In a previous study, we developed a test methodology for 131
132 the PTI to detect removed and damaged DPF [4]. This test 133
134 consists of measuring the PN emissions of a vehicle during a 135
136 low idle test: gearbox in neutral position, engine warmed up 137
138 and at low idle speed (i.e. without depressing the accelerator 139
140 pedal). The PN value (in $\#/cm^3$) is the result of 3 measure- 141
142 ments of 5 seconds each. 143

144 We analysed the efficacy of three different PN measure- 145
146 ment devices: the TSI NPET, the Pegasor Mi3 and the Testo 147
148 NanoMet3. These devices are already commercially avail- 149
150 able and are typically used for automotive applications (see 151
152 Table 1). The PN measurements were performed on every 153
154 EURO5 and EURO6 diesel passenger car using one of the 3 155
156 previously mentioned PN measurement devices. These mea- 157
158 surements were executed just after the opacity test which is 159
160 currently the only emission related test at the PTI for this 160
type of vehicles in Europe. The tests were performed either
by us or by the employees of different accredited PTI com-
panies across Belgium.

149 These 3 devices were compared on 68 vehicles before 149
150 being sent individually to PTI stations. These comparative 151
152 measurements were performed by introducing simultaneously 153
154 in the exhaust pipe the probe of the TSI device and the one 154
155 of either the Pegasor or the Testo. The results of the com- 155
156 parison showed a high correlation between the studied de- 156
157 vices. Some differences could be observed between the de- 157
158 vices and could be explained by the different measurement 158
159 principles (i.e. condensation particulate counting or diffus- 159
160 ing charging), the dilution ratio or the accuracy of the cut-off 160
value for the small particles (i.e. 23 nm). These differences
will be taken into account in the uncertainty of the results by

Brand	TSI	Pegasor	Testo
Model	NPET	Mi3	NanoMet3
Technology	Condensation Particle Counting	Diffusion Charging	Diffusion Charging
Measured sizes	23 nm to 1 μm	23 nm to 2.5 μm	23 nm to 700 nm
Measurement range [$\#/\text{cm}^3$]	1000 to 5e6	600 to 1.3e9	1e4 to 3e8
Removed volatiles	Catalytic Stripper (350°C)	Heated Sample Line (200°C)	Heated Sample Line (100°C) & Evaporation Tube (300°C)
Dilution	10:1	No	Variable (10:1 up to 300:1)

Table 1

Main technical characteristics of the 3 PN measurement devices.

161 integrating the prediction error of the correlation into the PN
 162 EF calculations. Indeed, the uncertainty of these PN mea-
 163 surement devices is typically below 50% [5], which is sig-
 164 nificantly below the correlation uncertainty (factor between
 165 3 and 4, see section 2.2). An example of a time series data
 166 comparing the TSI NPET and the Pegasor Mi3 is provided
 167 in Figure 1, which also shows the stabilization time of these
 168 2 devices when inserting the probe in the exhaust pipe.

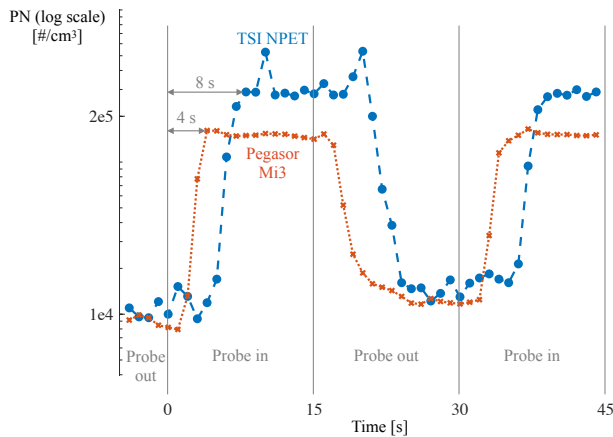


Figure 1: "Probe in" and "Probe out" correspond to the periods during which the probe was measuring inside the exhaust pipe and in the ambient air, respectively. The Pegasor Mi3 is the quickest device with a transformation time between 4 and 6 seconds, while the TSI NPET needs between 7 and 9 seconds to reach stable values.

169 The TSI NPET and the Testo NanoMet 3 were also com-
 170 pared to a AVL M.O.V.E PN PEMS iS device, fulfilling the
 171 Real Driving Emission requirements (see Figure 2). The
 172 results show a high correlation between this Real Driving
 173 Emissions (RDE) compliant device and these 2 devices used
 174 to measure PN at low idle at the PTI.

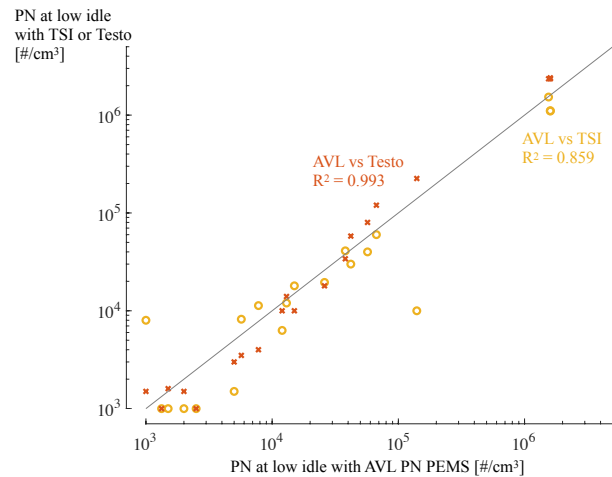


Figure 2: These comparative measurements show that the TSI NPET and the Pegasor Mi3 provide similar results compared the AVL PN PEMS, fulfilling the RDE requirements.

2.2. Determining the PN emission factor

The PN value measured at low idle ($\text{PN}_{\text{low idle}}$, expressed in $\#/\text{cm}^3$) is correlated to the PN emission factor (PN EF, expressed in $\#/\text{km}$) that would be obtained during laboratory test cycles such as the NEDC or the WLTC (see Figure 3). On this Figure, each point represents the test of one vehicle, which includes the PN at low idle test and the PN EF measured during a test. These two driving cycles allow to test the vehicles at different speeds and vehicles loads. To establish this correlation, we combined emission measurements performed during a low idle test and during an homologation cycle. This data comes from the JRC [8] and from TNO [10]. The high emitting vehicles were obtained by removing the DPF or by intentionally reducing its filtration efficiency. The idea of a correlation between low idle and cycle results were introduced in several publications of TNO and the JRC [8, 10]. There is no proof of causality but this correlation was shown as valid for many tests. Its principle is based on

193 a rather passive behaviour of the filter and a scaling effect
 194 between low idle and cycle conditions. Of course, the main
 195 idea is not to use this low idle measurement as an homologia-
 196 tion tool but rather to explore the ranges of emission factors
 197 on a large set of vehicles. The linear regression was obtained
 198 by considering only the measurements that have PN values
 199 at low idle above 10 000 #/cm³ because of the low accuracy
 200 of the diffusion charging measurement devices for very low
 201 concentrations (see Figure 3). This linear regression is thus
 202 only valid for the PN at low idle above 10 000 #/cm³, which
 203 corresponds to a PN EF of 1.559 × 10¹¹ #/km. Also, from
 204 the experimental data, it seems that this correlation becomes
 205 non-linear in the region below 10 000 #/cm³. Regarding the
 206 high emitters, the PN EF obtained from this correlation has
 207 been limited to 5.9 × 10¹³ #/km (it corresponds to a PN at low
 208 idle of 8.4 × 10⁶ #/cm³ by using the correlation), which cor-
 209 responds to the maximum value which was used for the cor-
 210 relation and which corresponds to a WLTC cycle of a vehicle
 211 without DPF. The grey area represents the 95% confidence
 212 interval of the regression which is computed by assuming
 213 that the prediction error follows a t-student distribution with
 214 15 degrees of freedom¹. This prediction error combines the
 215 measurement error associated to the PN at low idle (typically
 216 below 50% for these PN measurement devices) and to the
 217 PN Emission Factor (EF) (typically below 50%) [5, 16, 7].
 218 The upper and lower limits of the 95% confidence interval
 219 are within a factor between 3 and 4 compared the regression
 220 line, depending on the position on the x-axis (i.e. PN at low
 221 idle).

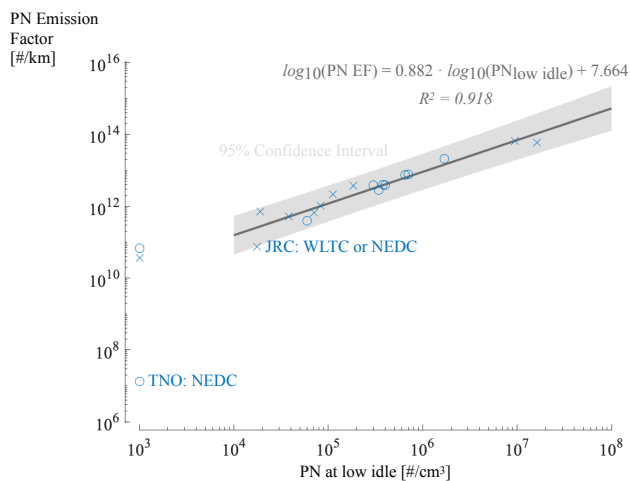


Figure 3: The PN EF for EURO5 and EURO6 diesel vehicles, that would be measured during NEDC or WLTC homologation cycle, can be estimated by a PN measurement at low idle. TNO data from [10] and JRC data from [8].

Thanks to the correlation between the PN low idle test (#/cm³) and the PN EF (#/km), the PN at low idle value of each tested vehicle will be converted into PN EFs using the

¹15 degrees of freedom because 17 x-y coordinates were used for the regression minus 2 degrees of freedom that were lost to estimate the slope and the y-intercept of the regression

following equation:

$$\log_{10}(\text{PN EF}) = 0.882 \cdot \log_{10}(\text{PN}_{\text{low idle}}) + 7.664,$$

for 10000 < PN_{low idle} < 8.4 × 10⁶ #/cm³. (1)

From these measurements PN at low idle test converted into PN EF, we have built an EF distribution for each emission standard, i.e. EURO5a, EURO5b and EURO6 diesel passenger cars. These distributions have also been further split by mileage categories, to analyse the degradation of emission performances of the fleet due to this factor.

3. Results and discussion

3.1. Measurements from the PTI

During this measurement campaign, 757 diesel passenger cars were measured, among which 629 are homologated according to the EURO5 emission standard (368 EURO5a and 261 EURO5b) and 128 according to EURO6. The split between EURO5a and EURO5b is due to the fact that the DPF was standard since the EURO5a emission regulations but became mandatory with the introduction of the EURO5b because of its limit on the PN emissions. The low amount of EURO6 vehicles can be explained by the more recent introduction of this emission standard, in 2014 for new type approval, and by the fact that vehicles usually undergo their first vehicle inspection after 4 years. The tested vehicles that are younger than 4 years either came because they are obliged to pass the PTI before being sold or because they are used professionally to transport people (e.g. taxis), in which case the PTI needs to be performed every 6 months. The mileage and age distribution of the tested EURO5a, EURO5b and EURO6 vehicles are provided in Tables 2, 3 and 4, respectively.

The PN at low idle values of these 757 EURO5 and EURO6 vehicles, measured by ourselves or by the employees of the PTI, are provided in Figure 4. This figure shows these measurement values as a Cumulative Distribution Function (CDF) to better understand what proportion of the fleet is below or above certain values. It shows that 53% of the vehicles have emissions below 5000 #/cm³, which corresponds to the PN concentration that is typically observed in ambient air at the different PTI stations. Also, 65% of the vehicles have emissions below 10000 #/cm³ which corresponds to the lower range of the Testo NanoMet3 and to the lower limit for the validity of the correlation between the PN at low idle and the PN EF. Finally, this figure also shows that 15% of the vehicles have PN emissions above 250 000 #/cm³ which is expected to be the future threshold to pass the test that will be implemented at the PTI in the coming years. This Figure also provides this information per EURO class.

3.2. PN emission factors distributions

Thanks to the existing correlation between the PN values measured at low idle and the PN EF (see Equation 1), the PN EF for each vehicle can be computed and then, using these individual PN EF values, the experimental CDF of

Age [years]	Mileage [thousands of km]				Total
	0-50	50-100	100-150	>150	
5-6.5	3	22	30	19	74
6.5-8	11	58	79	64	212
8-10	7	13	22	40	82
Total	21	93	131	123	368

Table 2
Mileage and age distribution for EURO5a vehicles.

Age [years]	Mileage [thousands of km]				Total
	0-50	50-100	100-150	>150	
2-4	17	65	20	7	109
4-6	8	54	48	27	137
6-8.5	0	4	5	6	15
Total	25	123	73	40	261

Table 3
Mileage and age distribution for EURO5b vehicles.

271 the tested fleet can be obtained (see Figure 5). The bands
272 around the experimental CDF correspond to the 95% con-
273 fidence interval and combines 2 effects. Firstly, the predic-
274 tion error when converting the PN at low idle into the PN
275 EF which is assumed to follow a t-student distribution with
276 15 degrees of freedom. Secondly, the bootstrapping method
277 is used to take into account the effects of the limited sample
278 size as well as the influence of the selection of the samples
279 among the entire population [2]. Since the regression was
280 computed with the PN at low idle values above 10000 \#/cm^3 ,
281 only the PN EF greater than the value corresponding to this
282 lower limit, i.e. $1.559 \times 10^{11} \text{ \#/km}$, were considered. The
283 PN EF below this limit represent the cleanest vehicles of the
284 fleet and although they fortunately represent 65% of the ve-
285 hicles, they only represent 0.8% of the PN emissions of the
286 tested fleet. In general, the EURO6 vehicles that were tested
287 have a lower PN EF than the EURO5 vehicles: 89% (confi-
288 dence interval between 85% and 94%) of the EURO6 vehi-
289 cles have PN emission factors below the homologation limit
290 (official limit when a vehicle is homologated according to the
291 EURO5b or EURO6 emission standard, i.e. $6 \times 10^{11} \text{ \#/km}$)
292 while only 85% (confidence interval between 79% and 88%)
293 of the EURO5b vehicles and 68% (confidence interval be-
294 tween 63% and 73%) of the EURO5a vehicles are below this
295 limit, assuming the correlation mentioned above.

On top of the experimental CDF of the EURO5a and EURO5b vehicles, the CDF of the lognormal distribution fitted to the data, using the Maximum Likelihood Estimation method, is also provided (see Figures 6 and 7). The lognormal distribution has the following probability density function, where μ is the mean and σ is the standard deviation:

$$f(x) = \frac{1}{x\sigma\sqrt{2\pi}} \exp\left(\frac{-(\ln x - \mu)^2}{2\sigma^2}\right).$$

296 The lognormal distribution is particularly well suited to rep-
297 resent actual fleet emissions, because it can only consider
298 positive values. Also, it is skewed with higher probabilities

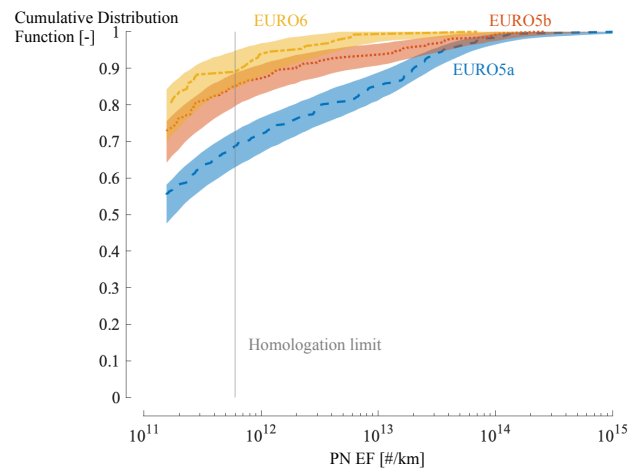


Figure 5: Experimental cumulative distribution functions of the computed PN EF (\#/km) for the EURO5a, EURO5b and EURO6 diesel passenger cars and their 95% confidence intervals.

for values lower than the average but with a long tail for the
299 high values representing the high emitters [12]. Additionally,
300 3 important PN EF are shown: the experimental arithmetic
301 average of the PN EF, the homologation limit (i.e. $6 \times$
302 10^{11} \#/km) and the arithmetic average HBEFA (Handbook
303 Emission Factors for Road Transport) value. For EURO5a,
304 there is no homologation limit so the value is indicative,
305 while the HBEFA does not make any difference between
306 EURO5a and EURO5b. The HBEFA provides emission factors
307 for the CO_2 emission, for the regulated pollutants but
308 also for some unregulated pollutants. These emission factors
309 are available for a wide range of vehicles categories,
310 fuels and EURO standards and consider different driving
311 conditions (urban, rural and highway) [11]. Initially, these
312 HBEFA emission factors were mainly based on measure-
313 ments from homologation cycles but now they include more
314

315 and more real-driving emission measurements with PEMS
 316 or Remote Sensing to better represent the reality. The HBEFA
 317 emission factors are commonly used for emission inventories
 318 to assess the emissions of the road transport sector.

319 In the case of EURO5a and EURO5b vehicles, the average
 320 HBEFA emission factor (arithmetic average of the urban,
 321 rural and highway emission factors based on equal distance
 322 shares, similar to the RDE regulations) equals 1.04×10^{12}
 323 $\#/km$ and is only slightly higher than the homologation
 324 figure, i.e. 6×10^{11} $\#/km$ (see Table 5). Assuming the previously
 325 mentioned correlation, the experimental arithmetic
 326 average based on the 368 EURO5a tested vehicles shows that
 327 the high emitters have a significant impact on the fleet arithmetic
 328 average emission factor, i.e. 5.79×10^{12} $\#/km$. Therefore,
 329 the homologation (not applicable for EURO5a) and the HBEFA
 330 figures are lower than the calculated fleet emissions (estimated
 331 based on low idle measurement and the correlation) for EURO5a
 332 by a factor 9.6 and 5.6, respectively. For the 261 EURO5b
 333 vehicles, the fleet arithmetic average emission factor equals
 334 2.60×10^{12} $\#/km$, which is higher by a factor 4.3 than to
 335 the homologation limit and by a factor 2.5 than the HBEFA
 336 figures. It should be highlighted that the homologation limit
 337 needs to be fulfilled for a specific driving cycle and under
 338 well defined conditions, which can explain the discrepancies
 339 with real-world driving and with the estimated PN EF.
 340

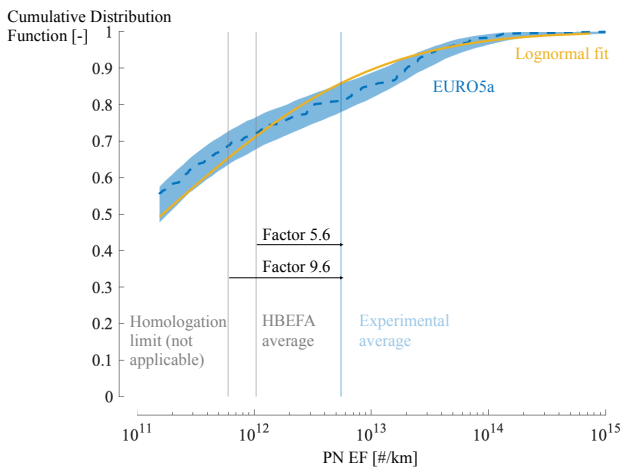


Figure 6: Cumulative distribution function of the computed PN EF ($\#/km$) for the EURO5a diesel passenger cars. Lognormal fit parameters: $\mu = 25.832$, $\sigma = 3.263$.

341 Given the number of EURO5 vehicles that were tested,
 342 the impact of mileage can be analyzed by comparing the PN
 343 EF CDF of 4 mileage categories: below 50 000 km, between
 344 50 000 and 100 000 km, between 100 000 and 150 000 km
 345 and finally above 150 000 km (see Figure 8). For this analysis,
 346 EURO5a and EURO5b vehicles will be combined into
 347 EURO5, since the goal is to compare the experimental data
 348 to the HBEFA degradation factor and increase the number of
 349 data per mileage category. The error bars represent the 95%
 350 confidence interval. Even though the confidence intervals
 351 are larger compared to those of the entire EURO5 fleet (due

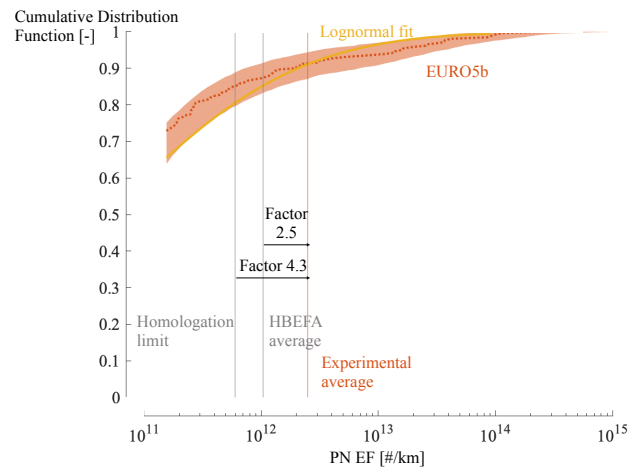


Figure 7: Cumulative distribution function of the computed PN EF ($\#/km$) for the EURO5b diesel passenger cars. Lognormal fit parameters: $\mu = 24.598$, $\sigma = 2.932$.

	Particle Number Emission Factor [$\#/km$]	
	EURO5a	EURO5b
HBEFA Urban	1.38×10^{12}	
HBEFA Rural	9.30×10^{11}	
HBEFA Motorway	8.12×10^{11}	
HBEFA Average	1.04×10^{12}	
Homologation limit	NA	6×10^{11}
Experimental	5.79×10^{12}	2.60×10^{12}

Table 5

HBEFA PN EF [14], homologation limit and experimental arithmetic average for the EURO5a and EURO5b diesel vehicles.

352 to the lower number of vehicles per mileage category), it can
 353 be observed that mileage has a significant impact on the PN
 354 EF: the arithmetic average emission factor for vehicles having
 355 a mileage higher than 150 000 km is 4.4 times higher
 356 compared to vehicles with a mileage below 50 000 km.

357 On top of the base emission factor that were discussed
 358 above, the HBEFA also provides emission factors that take
 359 into account the impact of mileage (up to 250 000 km), related
 360 to the German fleet in 2018. Considering this mileage
 361 degradation, the original PN EF (provided in Table 5) are
 362 increased by 0.6%, 5.1% and 10.2% for the urban, rural and
 363 highway driving conditions respectively, which results in an
 364 increase of 4.4% for the arithmetic average PN EF [14].
 365 Considering that the tested vehicles with a mileage below 50 000
 366 km represent the HBEFA figure without degradation, the PN
 367 EF of these low-mileage vehicles (i.e. 1.62×10^{12} $\#/km$)
 368 is very close to the HBEFA figure (i.e. 1.04×10^{12} $\#/km$).
 369 On the other hand, when considering the entire fleet which
 370 includes some very high emitting vehicles, the experimental
 371 arithmetic average reaches 4.47×10^{12} $\#/km$ while the
 372 HBEFA PN EF considering mileage degradation is only equal
 373 to 1.09×10^{12} $\#/km$. Based on these observations, it seems
 374 that the degradation of the emission performance is strongly

375 underestimated by the HBEFA emission factors.

376 A similar analysis was not made for EURO6 vehicles
 377 since only 128 vehicles were tested, 74% of them having a
 378 mileage below 100 000 km. Nevertheless, the PN at low idle
 379 measurements for EURO6 vehicles, together with EURO5a
 380 and EURO5b, are shown in Figures 9 and 10 with respect to
 381 age and mileage.

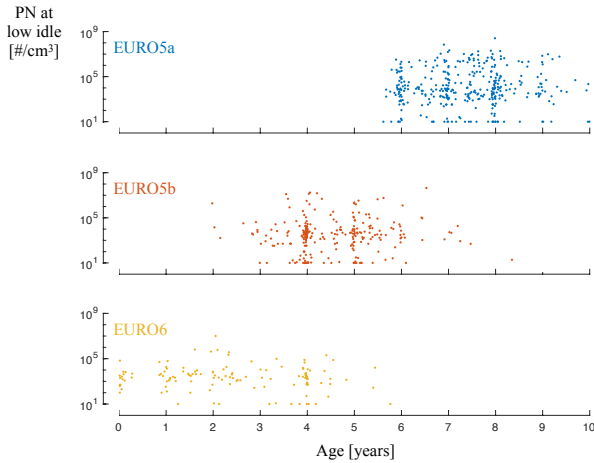


Figure 9: PN at low idle as a function of age for EURO5a, EURO5b and EURO6 vehicles.

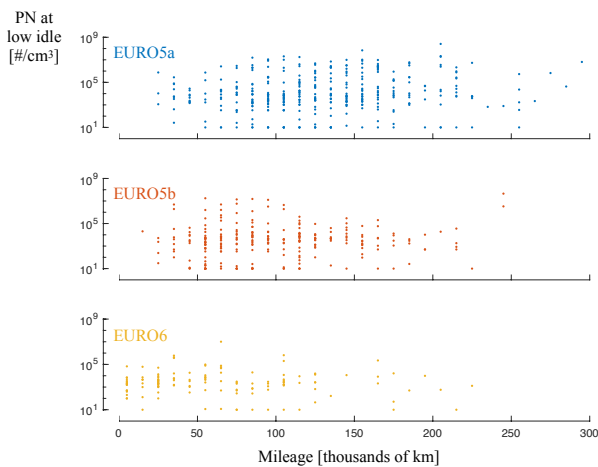


Figure 10: PN at low idle as a function of mileage for EURO5a, EURO5b and EURO6 vehicles.

382 Regarding the EURO6 vehicles, the CDF of the PN EF is
 383 shifted upwards compared to the EURO5 vehicles, meaning
 384 that this whole group of vehicles is characterized by lower
 385 emissions (see Figure 11). This can mainly be explained
 386 by the fact that EURO5 vehicles are older and have higher
 387 mileage which increases their emissions. Also, engine-out
 388 PN emissions might have been reduced for EURO6 vehicles
 389 together with improvements in DPF design and manufactur-
 390 ing processes. In the case of EURO6 diesel vehicles, the
 391 arithmetic average HBEFA PN EF (i.e. 1.39×10^{11} #/km)
 392 is 77% below the homologation limit (i.e. 6×10^{11} #/km,
 393 see Table 6). The gaps between the estimated PN EF (i.e.

	Particle Number Emission Factor [# /km]
HBEFA Urban	1.36×10^{11}
HBEFA Rural	1.03×10^{11}
HBEFA Motorway	1.77×10^{11}
HBEFA Average	1.39×10^{11}
Homologation limit	6×10^{11}
Experimental	7.58×10^{11}

Table 6

HBEFA PN EF for EURO6a and EURO6b vehicles (123 out of the 128 EURO6 vehicles are EURO6a or EURO6b) [14], homologation limit and experimental average for the EURO6 diesel vehicles.

394 7.58×10^{11} #/km, computed from low idle PN measure-
 395 ments using the correlation) and the HBEFA and homologa-
 396 tion figures are reduced compared to EURO5: the HBEFA
 397 and the homologation figures underestimate the estimated
 398 PN EF by factors equal to 5.5 and 1.3, respectively. In the
 399 case of EURO6 vehicles, it is difficult to analyze the effect
 400 of mileage due to the lower number of tested vehicles (128
 401 EURO6 compared to 629 EURO5 vehicles).

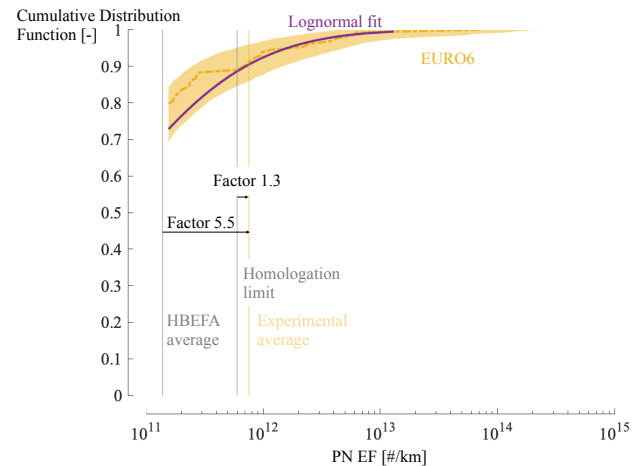


Figure 11: Cumulative distribution function of the computed PN EF (#/km) for the EURO6 diesel passenger cars. Lognormal fit parameters: $\mu = 24.406$, $\sigma = 2.250$.

4. Conclusion

402 This work provides PN emission factor distributions for
 403 EURO5a, EURO5b and EURO6 diesel passenger cars based
 404 on the measurements of 757 vehicles. The measurements
 405 performed on this large fleet consist of measuring the PN
 406 concentration at low idle, which shows high correlation with
 407 PN emission factors that would be obtained during expen-
 408 sive and time consuming homologation cycles. The vehi-
 409 cles were tested during their Periodic Technical Inspection,
 410 which allows to collect additional information such as the
 411 mileage of the vehicle, information that is not always avail-
 412 able when performing remote sensing measurements for ex-
 413 ample.
 414

415 These distributions show that the majority of the vehi- 473
 416 cles are clean and have emission factors, estimated from the 474
 417 PN at low idle measurement and the correlation, below the 475
 418 homologation limit (introduced with the EURO5b emission 476
 419 standard) of 6×10^{11} #/km (i.e. 68% of the EURO5a, 85% of 477
 420 the EURO5b and 89% of the EURO6 vehicles). Neverthe- 478
 421 less, they also show the presence of high emitters of partic- 479
 422 ulate matter which have a dramatic impact on the arithmetic 480
 423 average PN emission factors of the fleet. For the EURO5a 481
 424 vehicles, the high emitters induce a severe increase of the ob- 482
 425 served experimental arithmetic average (i.e. 5.79×10^{12} #/km), 483
 426 estimated from the PN at low idle measurements, which is 484
 427 higher by a factor 5.6 compared to the HBEFA figure. This 485
 428 can be explained by the mileage degradation effect and by 486
 429 the presence of vehicles with very high mileage within this 487
 430 vehicle category. For the EURO5b vehicles, the estimated 488
 431 PN EF (i.e. 2.60×10^{12} #/km) is higher than the homologa- 489
 432 tion limit and the HBEFA figure by a factor 4.3 and 2.5, re- 490
 433 spectively. Regarding the EURO6 vehicles, the experimen- 491
 434 tal arithmetic average (i.e. 7.58×10^{11} #/km) is higher by a 492
 435 factor 1.3 compared to the homologation limit and by a fac- 493
 436 tor 5.5 compared to the HBEFA figure. Because of the sig- 494
 437 nificant impact of the high emitters, the sampling of the ve- 495
 438 hicles is of utmost importance and a large fleet is required to 496
 439 increase the representativeness of the sample and to capture 497
 440 the effect of the high emitters. Therefore, more EURO6 ve- 498
 441 hicles should be tested to validate the obtained distributions. 499

442 Using distributions to characterize the PN emission fac- 500
 443 tors allows to clearly observe the presence of high emitters 501
 444 among the fleet, and removing these high emitters should be 502
 445 the priority when it comes to reducing the emissions of a 503
 446 certain fleet. Also, these distributions and their confidence 504
 447 intervals could be used as input for emission inventories to 505
 448 assess more carefully the impact of the road transport sector 506
 449 for a specific region, by taking the uncertainties into account. 507
 450 Indeed, the emission factors for emission inventories could 508
 451 be modelled as lognormal distributions, which represent the 509
 452 experimental distributions with high fidelity. 510

453 Future work could collect data from a larger number of 511
 454 EURO6 vehicles to analyse the influence of mileage for this 512
 455 emission standard. Also, the fitted log-normal distributions 513
 456 could be tuned to match a specific national fleet and then 514
 457 used in the emission models to take into account the vari- 515
 458 ability of the PN emission factors among the fleet. 516

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 access to vehicle repair and maintenance information, amending direc- 480
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Age [years]	Mileage [thousands of km]				Total
	0-50	50-100	100-150	>150	
0-2	45	5	2	1	53
2-4	12	26	13	7	58
4-6	1	6	7	3	17
Total	58	37	22	11	128

Table 4
Mileage and age distribution for EURO6 vehicles.

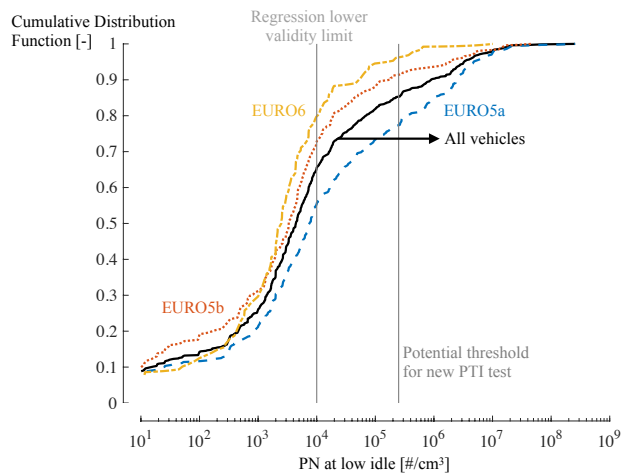


Figure 4: Cumulative distribution function of the PN measurements performed at low idle ($\#/cm^3$) of the all the tested vehicles but also for each individual EURO class.

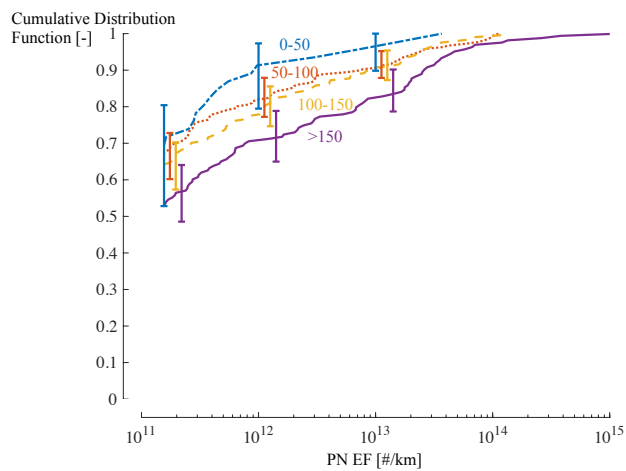


Figure 8: Cumulative distribution function of the computed PN EF ($\#/km$) for the EURO5 diesel passenger cars for 4 different mileage categories. The mileage categories are identical to the ones used in Tables 2 and 3 and the legends on the graph are expressed in thousands of kilometers.