

This is a repository copy of *Measuring the success of reducing emissions using an on-board eco-driving feedback tool.* 

White Rose Research Online URL for this paper: http://eprints.whiterose.ac.uk/101829/

Version: Accepted Version

## Article:

Caulfield, B, Brazil, W, Ni Fitzgerald, K et al. (1 more author) (2014) Measuring the success of reducing emissions using an on-board eco-driving feedback tool. Transportation Research Part D: Transport and Environment, 32. pp. 253-262. ISSN 1361-9209

https://doi.org/10.1016/j.trd.2014.08.011

© 2014, Elsevier. Licensed under the Creative Commons Attribution-NonCommercial-NoDerivatives 4.0 International http://creativecommons.org/licenses/by-nc-nd/4.0/

#### Reuse

Unless indicated otherwise, fulltext items are protected by copyright with all rights reserved. The copyright exception in section 29 of the Copyright, Designs and Patents Act 1988 allows the making of a single copy solely for the purpose of non-commercial research or private study within the limits of fair dealing. The publisher or other rights-holder may allow further reproduction and re-use of this version - refer to the White Rose Research Online record for this item. Where records identify the publisher as the copyright holder, users can verify any specific terms of use on the publisher's website.

#### Takedown

If you consider content in White Rose Research Online to be in breach of UK law, please notify us by emailing eprints@whiterose.ac.uk including the URL of the record and the reason for the withdrawal request.



eprints@whiterose.ac.uk https://eprints.whiterose.ac.uk/

# Measuring the success of reducing emissions using an on-board eco-driving feedback tool

Brian Caulfield<sup>\*</sup>, William Brazil, Kristian Ni Fitzgerald and Craig Morton

Department of Civil, Structural and Environmental Engineering, Trinity College Dublin, Ireland

#### Abstract

This paper reports the findings of an eco-driving trial that was designed enable users to make pre-trip and on-route decisions when driving as to the optimal route to take. The basis of this paper will be to estimate how efficiently drivers are performing in relation to fuel consumption per kilometers (KM). The analysis uses details on the vehicle specification, in terms of fuel efficiency, and relates this to the distance travelled to provide the user with information on the efficiency per KM travelled. Eco-driving involves the training of individuals to change their driving patterns and to adapt to driving conditions. The results of the study show that eco-driving feedback is a powerful tool and how it can be used to reduce emissions.

#### 1. Introduction

In recent years many authors have written about the success of eco-driving and its ability to reduce emissions. Barkenbus (2010) suggests that eco-driving is the overlooked climate change initiative and that following a policy of eco-driving can result in a 10% reduction in fuel consumption which will have a knock on effect of reducing emissions. A range of studies have shown that the benefits from eco-driving can range from a 5 to 20% reduction in emissions (Stillwater et al, 2012).

Beusen et al (2009) examined 10 cars over a 10-month period after taking a course, which provided them with eco-driving training. The authors found that drivers on average had a 5.8% reduction in fuel usage. However, the study showed that the fuel savings deminished over time and drivers went back to their original habits. Delhomme et al (2013) conducted a survey of French drivers to ascertain their opinions in relation to eco-driving and how they feel about adopting eco-driving styles. The findings show that generally respondents said it would be easy to adapt to the eco-driving styles. The results also found that younger and middle aged drivers said it may be difficult to adapt to the driving styles.

Boriboonsomsin et al (2011) conducted a study of 20 drivers in Southern California using an on-board eco-driving feedback tool. The findings of the study showed modest increases in fuel economy of 6% for urban streets and 1% on motorways. This was attributed increased congestion in the area. Martin et al (2013) conducted a study of 18 drivers in California using on-board feedback for eco-driving. The study took a similar approach to the one reported in this paper in that the devices were turned off for the first month and then switched on to give drivers feedback on driving style. Similar to the results found in Boriboonsomsin et al (2011), the authors show that modest improvements in fuel efficiency. In 2012, Martin et al (2012) conducted a longitudinal study of a sample of participants in California. This study surveyed participants over three time intervals to determine if eco-driving behavior

<sup>\*</sup> Corresponding Author: Email: brian.caulfield@tcd.ie; Tel: +353 1 896 2534

would last in the long run using information from an eco-driving website. The study looked at before and after information on how the study worked. The results showed that more than half of the sample improved their eco-driving behavior and that females, those living in smaller households and those with newer cars were more likely to improve eco-driving behavior.

Stillwater and Kurani (2012) employed the theory of planned behavior to examine how driving behaviors change using an on-board eco-driving feedback tool. The findings showed that that setting goals for participants and real-time feedback resulted in drivers increasing their fuel efficiency. Rutty et al (2013) examined the impacts of eco-driving on Calgary's municipal fleet. In the study fifteen drivers in a study to reduce the emissions associated with vehicle idling. The results of the study showed that average vehicle idling was reduced by between 4% and 10% per day. Other road users have been examined to ascertain if eco-driving can be applied to public transport drivers. Sromberg and Karlsson (2013) examined bus drivers in Sweden using in vehicle feedback tools to reduce harsh acceleration. The findings of the study showed that a 6.8% reduction in fuel usage occurred in the study period.

The research presented in this paper seeks to examines the potenial of ecodriving over a ten month period. Most of the other studied in this field have taken place over a shorter period of time and few have the high number of participats that this study used. Therefore previous studied have been unable to have been unable to examine a large sample over such a long period of time. The purpose of the research was to determine which factors and technologies have the greatest impacts on the potential emissions savings. The research presented adds to the body of work conducted in this field as it shows the potential differences between several technologies.

#### 2. Methodology and data collection

### 2.1 Information provided in the trial

The data collected for this study was collected using fully instrumented vehicles and some groups of respondents were provided with real-time eco-driving feedback using an on-board satnav device. Figure 1 shows a screen shot of the information provided to users of the on-board feedback. As part of the active driver feedback the driver was provided with speeding alerts, excessive maneuvers when cornering and braking, idling alerts and real-time fuel consumption information when driving. Drivers were also given traffic information and if routes were congested alternative routes were suggested.

Figure 2 shows the device that was used during the trial a TomTom ecoPLUS<sup>TM</sup> navigation device. The on-board feed back, from the TomTom ecoPLUS<sup>TM</sup>, provided users with advice in relation to their driving style and routes to minimize emissions. The 'WEBFLEET' website provided feedback to the participants in relation to their emissions and suggestions to reduce emissions.



Figure 1 Information provided to the driver



Figure 2 TomTom ecoPLUS<sup>TM</sup>

During the trial the following information was collected:

- Fuel consumption per liter/100 km
- Idling time
- How much km is driven

## 2.2 Data Collection and trial

The eco-driving trial started in January 2012, in the Netherlands, and the results presented in this paper track the vehicles up to October 2012. In total 167 participants took part in the trial driving an instrumented Peugeot 107 XS 1.0. These participants were selected after a recruitment drive that advertised for participants in the trial with a description of the main objectives of the trial.

Five different groups were analyzed during the trial period. Participants drove on a mixture of roads and in both urban and rural situations.

**Group A:** This group had 82 users and these users were provided with on-board active driver feedback for the duration of the trial and access to WEBFLEET online.

**Group B:** This group had 27 users and these users were provided with on-board active driver feedback for the duration of the trial and from the  $1^{st}$  of July 2012 were given access to WEBFLEET online.

**Group C:** This group had 27 users and for the first two months had no interventions. Then this group was given both on-board active driver feedback and WEBFLEET online.

**Group D:** This group had 16 users and was not given any on-board information. This group was given WEBFLEET online from March 2012.

**Group E:** This group had 15 users and they received no information at all on driving style. This group was used a reference group to compare the other groups.

While it would have been ideal to have equal numbers in each of the five groups for the analysis, the decision to divide the sample into the proportions shown above was outside of the control of the authors. To adapt to the different sizes in the samples the analysis presented in this paper is only conducted within groups and then comparisons are made between these groups. Also ideally it would have been better to have more participants in Group E as the control group – but this was outside of the control of the authors of this paper.

Table 1 presents a description of the participants contained within the five user groups in terms of their personal characteristics alongside a number of driving related attributes. This has been tabulated in order to determine the degree to which the groups are comparable in reference to their basic features. For all of the variables included in Table 1, appropriate hypothesis test have been employed to determine if the identified differences between groups are significant.

Examining the outcome of this analysis, it is apparent that the user groups are significantly different from each other for all of the characteristics inspected. Reviewing the personal characteristics of the groups, it is evident that Group D and Group E tend to contain younger participants whilst Group C is populated by older individuals. The gender split between the groups is reasonable similar, though Group D has a higher prevalence of males. In terms of environmental concerns, Group D stands out with a high proportion of participants who are concerned about the environment, which may partially explain why participants of Group D are also the most likely to describe their driving style as economic. Turning the attention to the driving related attributes, years of driving experience tends to follow the variation observed in the age profiles of the groups whilst there are no large differences observed between groups in reference to the distance they drive per trip and the time they spend idling per mile. A larger degree of group separation is identified in terms of time spent speeding, with Group F exhibiting a relatively large average for this characteristic.

Reflecting on the results of these user group comparisons, it is evident that the groups used in this project display a number of visible differences from each other in terms of their personal characteristics and driving profiles. This has likely been generated as a result of the constrained nature of the project, which was unable to follow a strictly random sampling procedure. With this in mind, user group comparisons in the following analysis should be interpreted with caution. However, the user group differences identified in this study are unlikely to significantly affect how participant's behaviour changed over time meaning that temporal variations in the results, such as the degree to which a participant's fuel consumption decreased, are unlikely to be biased.

		User Group						
		Α	В	С	D	Ε		
Age*	18-30	21.8%	8.0%	7.8%	27.2%	48.6%		
	31-50	43.8%	54.2%	35.8%	55.3%	32.2%		
	51-65	24.6%	22.1%	49.1%	17.5%	19.2%		
	65+	9.8%	15.7%	7.3%	0.0%	0.0%		
Gender*	Male	60.8%	67.7%	69.0%	72.0%	68.5%		
	Female	39.2%	32.3%	31.0%	28.0%	31.5%		
Environmental	Unconcerned	4.3%	24.2%	0.0%	0.0%	0.0%		
Concern*	Indifferent	40.3%	25.5%	42.2%	18.3%	54.0%		
	Concerned	55.4%	50.3%	57.8%	81.7%	46.0%		
Driving Style*	Sporty	28.6%	21.1%	27.5%	20.1%	24.4%		
	Economic	14.6%	21.0%	13.7%	49.9%	21.3%		
	Safe	56.7%	58.0%	58.8%	30.0%	54.2%		
Driving	1 to 5 years	10%	11.4%	4.2%	9.6%	27%		
Experience*	5 to 10 years	14.8%	12.5%	3.6%	17.6%	26.6%		
	10 to 20 years	20.2%	23.1%	17.7%	33.2%	14.5%		
	20 years+	55%	53.1%	74.4%	39.6%	31.8%		
Distance Per Trip	Mean	19.13	21.65	17.65	22.73	16.27		
(km)*	S.D	24.78	28.30	21.57	38.27	27.24		
Idle Time (per km)*	Mean	.452	.237	.377	.475	.544		
	S.D	2.355	1.197	1.354	2.463	3.872		
Speeding Time*	Mean	199.23	270.61	280.16	316.01	383.85		
	S.D	282.05	437.92	366.70	379.84	383.85		

Table 1 Description of user groups employed in the project

\* Between group differences valid at the .000 level

#### 3. Results and analysis

This section of the paper presents the analysis conducted on the data collected in the trial.

#### **3.1 Testing for Normality**

Initially the data was examined to determine whether or not it was normal or not to determine the best approach with regard to analysing data. Figure 3 indicates that the data is highly non-normal in distribution. As the data is non-normal in nature, a number of standard statistical test are no longer deemed valid. To overcome this problem, a number of non-parametric tests were conducted

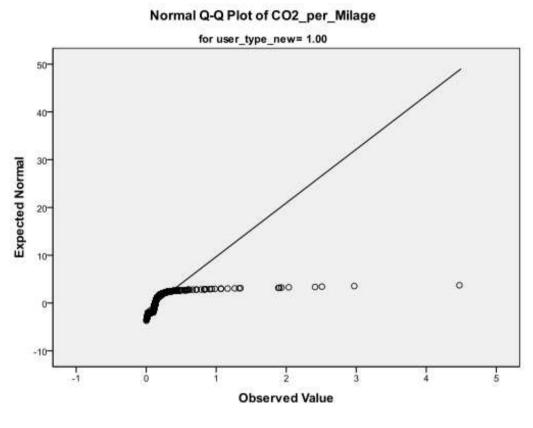


Figure 3 Tests for normality

#### **3.2 Descriptive Statistics**

Table 2 presents a description of the trip characteristics of those participants in each of the groups in the trial. Table 2 presents the average of the sample and the standard deviation of the characteristics of the driving undertaken by each group. The results show that characteristics such as average daily travel distance and the number of trips are similar.

					-					
	Grou	ıp A	Grou	ıp B	Group C		Group D		Grou	рE
	Mean	S.D.	Mean	S.D.	Mean	S.D.	Mean	S.D.	Mean	S.D.
Number of trips	3.4	2.0	3.2	2.0	3.6	2.0	3.4	1.9	3.4	1.8
Distance (in KM)	56.8	64.6	61.2	72.8	57.3	66.5	69.0	79.7	49.0	63.1
Driving time (in mins)	59	50	62	56	59	51	69	60	55	47
Fuel usage (in liters)	3	3.4	3.5	4.2	3.2	3.7	3.7	4.5	2.8	3.4
Idle time (in mins)	9	8	12	11	13	16	13	15	22	18
Average time speeding (mins)	3	4	4	7	5	6	5	6	3	6

#### Table 2 Description of the data collected

## 3.3 Non Parametric Tests Between all Groups

For the purpose of analysis the primary metric under consideration is grams of carbon dioxide produced per kilometre driven. Table 3 displays the results of the non parametric test. Results of the Kruskal –Wallis test indicate that there are statistical differences between the observed means of the carbon dioxide emissions per kilometre, for the five experimental groups under examination.

	Table 3: Non	parametric te	est results	between	groups
--	--------------	---------------	-------------	---------	--------

Total N	22,231
Test Statistic	172.016
Degrees of freedom	4
Asymptotic Sig. (2 Sided test)	.000

#### 3.4 Emissions per group

Table 4 presents the means for each of the groups under examination. It can be observed that the highest average emissions are produced by the control group (Group E) and the lowest emissions were associated with Group D. The number of trips represents the total number of trips in each of the groups examined.

Group	Mean of CO2 emissions per KM	Number of Trips	Std. Deviation of CO2 emissions per KM
А	.1387	9,897	.0889
В	.1360	3,844	.0379
С	.1354	3,831	.0305
D	.1308	2,458	.0260
E	.1409	2,201	.0395
Total	.1370	22,231	.0645

 Table 4 Comparison of means (CO2 per KM)

As engines experience much greater emissions per km in the initial kilometres, due to the effects of start up emissions, it was decided to examine trips of varying distances. Table 5 presents all trips and their aver CO2 emissions per KM for trips 0-5 KM, 5-10 KM and all trips over 10KM. In all cases, as with unrestricted trips, the highest observed emissions are associated with the control group.

Group	Mean of CO2	8	Std. Deviation of CO2
	emissions per KM	Number of Trips	emissions per KM
0-5KM trips			
Α	.1336	9,778	.033
В	.1347	3,820	.027
С	.1347	3,812	.027
D	.1303	2,445	.025
E	.1398	2,190	.031
Total	.1342	22,045	.030
5-10KM trips			
A	.1296	9,136	.025
В	.1314	3,585	.022
С	.1315	3,479	.023
D	.1269	2,261	.019
Е	.1367	2,030	.027
Total	.1307	20,491	.024
Trips over 10KM			
Α	.1387	9,897	.089
В	.1360	3,844	.038
С	.1354	3,831	.031
D	.1308	2,458	.026
E	.1409	2,201	.040
Total	.1370	22,231	.065

Table 5 Comparison of means (CO2 per KM) 0 – 5 KM in length

As both groups C and D involved interventions, both coming after two months, it is interesting to compare the mean emissions of users in each of these groups and examine the percentage reduction that the interventions have created. These two groups were isolated as they were the only two that had an intervention after a period of operation. As there are issues regarding data collection, particularly in the first two months, on 9 and 7 respondents, for groups C and D respectively, have enough observations to conduct meaningful analysis. Results indicate that, for both user groups, meaningful and significant emissions reduction has occurred (as seen in Tables 6 and 7).

User	% Reduction
1	8.64
2	7.43
3	-12.10
4	10.88
5	10.26
6	14.56
7	2.15
8	10.34
9	27.49
Overall	8.85

Table 6 Group C Emissions Reduction

With regard to Group C average emissions reductions was observed to 8.85%, with the highest observed emissions reduction being 27.5%. The average emissions figure is somewhat skewed by participant number 3, who actually produced an emissions increase of 12%. If this participant is excluded from analysis the figure would be 11.2% Emission reductions for Group D were observed to be even lower than those observed for Group C, with an average emissions reduction of 10.3%.

User	% Reduction
1	4.850
2	20.842
3	10.690
4	11.193
5	3.227
6	7.425
7	14.157
Overall	10.341

Table 7 Group D: Emissions Reduction

## **3.5** Emissions over the time period of the trial

The following section reports the reductions in CO2 emissions from each of the five groups examined in the trial. In order to examine what if any reductions in CO2 occurred the emissions from the first two weeks of driving were averaged and used a baseline to compare subsequent weeks for reductions in emissions.

The results presented in Table 8 show the first two-week average and then the data is broken down into 5-week periods to show the changes in emissions over time. The results show that for Group A that there is a drop in emissions from week 13 to week 27 and then there is a rebound and the average emissions increases. The results for Group B show that there is a steady drop in the average emissions in the first 12 weeks of the trial. From week 13 to 22 the results show a greater decline in the average CO2 emissions per KM compared to the other time periods. This time period was when those in Group B were provided with information via WEBFLEET. The results for Group C also show a steady decline in the average CO2 emissions. This group was provided with on-board information and WEBFLEET from week 8. The results show that from week 8 there was a decrease in average emissions. The findings for Group D show that from week 12, when the participants got access to WEBFLEET, that a decrease in average emissions was experience by those in this group. Group E was used as the reference group in this study, as they were given no extra information on driving performance. As one would expect there was little change in the over all average CO2 emissions per KM in this group.

Weeks	Group	Group A		В	Group	Group C		D	Group I	
	Mean	S.D.	Mean	S.D.	Mean	S.D.	Mean	S.D.	Mean	S.D.
1&2	.133	.062	.138	.021	.144	.006	.144	.031	.142	.072
3-7	.135	.067	.135	.029	.137	.030	.137	.027	.138	.035
8-12	.136	.010	.133	.035	.135	.023	.134	.024	.132	.024
13-17	.130	.012	.130	.029	.133	.037	.131	.031	.139	.055
18-22	.131	.035	.130	.027	.134	.037	.128	.029	.138	.039
23-27	.130	.035	.133	.021	.135	.035	.128	.028	.142	.067
28-32	.136	.037	.134	.046	.135	.012	.125	.025	.138	.062
33-37	.136	.017	.133	.034	.132	.028	.128	.026	.141	.041

Table 8 Overall reductions in CO2 emissions per KM

In order to ascertain if drivers had different behavior on weekends compared to weekdays the dataset was divided between weekdays and weekends to determine if there was any difference. Table 9 presents the results for the weekends and Table 9 presents the results for weekdays. The findings for Group A shows that on average participants had higher average emissions on weekends. The results from Group B also show a similar trend with higher average emissions on weekends compared to weekdays. These trends are also shown for Groups C, D and E.

Table 9 Reductions in CO2 emissions - Weekends

	bic > Reductions in CO2 classions - Weekends									
Weeks	Group A	١	Group E	3	Group C		Group D		Group E	
	Mean	S.D.	Mean	S.D.	Mean	S.D.	Mean	S.D.	Mean	S.D.
1&2	.134	.054	.142	.068	.135	.043	.162	.173	.136	.010
3-7	.138	.042	.143	.073	.147	.092	.142	.143	.147	.049
8-12	.137	.043	.139	.053	.129	.117	.137	.044	.135	.025
13-17	.135	.021	.133	.050	.139	.044	.130	.045	.138	.087
18-22	.137	.037	.129	.020	.139	.015	.129	.073	.144	.061
23-27	.137	.049	.129	.064	.139	.059	.128	.018	.150	.231
28-32	.138	.029	.138	.064	.133	.066	.126	.076	.138	.077
33-37	.136	.006	.140	.057	.131	.049	.130	.018	.146	.083

#### Table 10 Reductions in CO2 emissions - Weekdays

Weeks	Group A Group B		Group C		Group D		Group E			
	Mean	S.D.	Mean	S.D.	Mean	S.D.	Mean	S.D.	Mean	S.D.
1&2	.119	.100	.134	.121	.134	.146	.133	.074	.143	.076
3-7	.133	.171	.135	.039	.137	.062	.138	.018	.136	.038
8-12	.136	.140	.132	.036	.136	.048	.138	.019	.131	.032
13-17	.125	.078	.129	.021	.132	.029	.132	.031	.138	.061
18-22	.123	.148	.124	.164	.134	.039	.127	.052	.137	.051
23-27	.120	.171	.120	.038	.134	.039	.128	.046	.138	.051
28-32	.136	.032	.136	.074	.134	.046	.125	.028	.136	.067
33-37	.135	.023	.139	.058	.132	.027	.127	.012	.139	.043

This section of the paper presents a comparison between the results found for each of the groups to determine which interventions had the greatest impact upon CO2 emissions per KM. Table 11 presents the findings of a comparison on the average weekly CO2 emissions per KM from groups A-D and compared against group E. This set of results shows how each of the test groups preforms against the control group. The results in Table 10, if positive show that the control group being compared had a reduction in CO2 emissions in the week in question compared to the control group. Whereas a negative result would indicate that that the group being

compared to the control group had higher average emissions per KM driven. The results in Table 10 show that on average each of the test groups had a greater reduction in CO2 emissions compared to the control sample. The results show that Group D performed the best with an average reduction in emissions of 6% compared to the control group. Groups A and B also had on average a 4% reduction in CO2 emissions compared to the control group with those in group C having a 3% decrease in emissions.

	Group A	Group B	Group C	Group D
Week 1	7%	1%	-5%	-3%
Week 2	8%	6%	2%	1%
Week 3	8%	3%	1%	2%
Week 4	3%	1%	0%	1%
Week 5	-7%	3%	2%	-1%
Week 6	3%	1%	0%	1%
Week 7	3%	2%	0%	0%
Week 8	-3%	-2%	-5%	-6%
Week 9	-2%	-5%	-2%	-2%
Week 10	-12%	1%	2%	1%
Week 11	3%	4%	-3%	-1%
Week 12	0%	0%	-2%	0%
Week 13	4%	6%	3%	2%
Week 14	7%	6%	8%	7%
Week 15	1%	4%	0%	-3%
Week 16	12%	10%	6%	14%
Week 17	10%	7%	6%	9%
Week 18	12%	13%	3%	14%
Week 19	7%	6%	1%	7%
Week 20	4%	4%	3%	3%
Week 21	4%	6%	7%	11%
Week 22	-1%	-1%	-2%	4%
Week 23	18%	18%	14%	22%
Week 24	5%	4%	-2%	11%
Week 25	14%	7%	8%	10%
Week 26	6%	4%	5%	8%
Week 27	2%	1%	0%	5%
Week 28	-7%	-3%	-4%	6%
Week 29	-3%	5%	-2%	4%
Week 30	3%	-2%	1%	10%
Week 31	7%	8%	10%	15%
Week 32	7%	8%	5%	14%
Week 33	6%	9%	13%	15%
Week 34	1%	1%	2%	6%
Week 35	5%	8%	6%	12%
Week 36	1%	0%	5%	11%
Week 37	7%	11%	6%	8%
Average reduction in CO2 emissions	4%	4%	3%	6%

Table 11: Groups compared to control group

#### **3.6 Socio Economic Data**

In order to explore how the driving related variables measured in this project vary in reference to the personal characteristics of the participants, a cross-tabulation has

been conducted with the results presented in Table 12. The driving related variables which are explored in this analysis are  $CO_2$  emissions per kilometre, kilometres per trip and the time spent speeding during a trip. Hypothesis testing has been utilised to determine if the differences identified are statistically significant. The results of these tests indicate that there is a substantial amount of variance identified, with all of the personal characteristics displaying significantly different averages for the driving related variables included.

In terms of  $CO_2$  emissions per kilometre, it is apparent that participants aged 65 years and older emit significantly more compared to participants in the other age ranges and that male participants emit larger quantities compared to female participants. Participants who stated they are unconcerned about the environment tend to emit higher levels of  $CO_2$  compared to those participants who are concerned, whilst participants who state they have a sporty driving style emit significantly higher  $CO_2$ levels than those who consider their driving to be economic. Whilst the difference observed in the  $CO_2$  emissions of participants of different levels of driving experience is significant, there does not appear to be any substantial variation on this characteristic.

Shifting the focus to kilometres per trip, a number of interesting differences have been identified. Participants aged 65 years or older tend to drive significantly shorter distances, perhaps indicating that their car use is more orientated around local casual journeys. Female participants tend to have shorter trip lengths compared to males whilst participants who stated they are unconcerned about the environment have longer trip lengths than concerned participants. In reference to participant driving style, individuals who stated they drive in an economic manner have longer trip lengths compared to participants who state they drive in a safe manner whereas participants with the fewest years of driving experience tend to have shorter trip lengths.

Whilst not specifically related to the direction of this paper, Table 12 also includes the average time a participant exceeded the speed limit. This variable has the potential to act as an indirect indicator of a participant's willingness to follow regulations which govern road use and may signify how receptive a participant would be to the introduction of new eco-driving protocols. In this instance, it is evident that younger participants who are aged between 18 and 30 tend to speed significantly more than older participants whilst males have a significantly higher average speeding time compared to females. Concerning the stated environmental concern and driving style of participants, significant differences in average speeding time are observed though the magnitude of these differences is not large. In terms of the level of driving experience of a participant, it is evident that those who have between 5 and 10 years experience tend to have high average speeding times than participants in other experiences ranges.

Table 22 Socio-economic results from the sample

		CO <sub>2</sub> per KM		KM per Trip		Speeding Time	
		Mean	S.D	Mean	S.D	Mean	S.D
Age	18-30	.1350**	.05127	20658**	23929	353.91**	531.99
	31-50	.1362**	.15724	19679**	24792	199.51**	243.32
	51-65	.1357**	.08273	21087**	32931	203.31**	276.35
	65+	.1569**	.05964	13631**	25526	126.19**	137.61
Gender	Male	.1403**	.13594	20328**	29486	251.44**	382.30
	Female	.1318**	.06003	18634**	22189	207.05**	274.93
Environmental Concern	Unconcerned	.1468**	.37939	21545**	28597	246.22**	256.07
	Indifferent	.1382**	.05262	19967**	27830	254.92**	368.65
	Concerned	.1360**	.07495	19533**	26782	219.99**	347.63
Driving Style	Sporty	.1428**	.20500	19930**	27584	252.56*	378.78
	Economic	.1263**	.05523	22008**	32064	213.15*	260.50
	Safe	.1398**	.07165	17885**	24008	237.80*	372.53
Driving Experience	1 to 5 years	.1349**	.06589	18342**	23009	241.04**	356.88
	5 to 10 years	.1387**	.03936	22767**	25169	422.55**	581.92
	10 to 20 years	.1333**	.06737	20739**	25238	206.03**	239.56
	20 years+	.1396**	.15239	19050**	29687	187.89**	253.40

\* Between group differences valid at the .00

\*\* Between group differences valid at the .000

With participants being asked as part of the project to state how concerned they are about environmental issues across three different response categories (unconcerned, indifferent, concerned), it proves interesting to examine the socioeconomic and driving characteristics of these three different participant groups. Table 13 presents an overview of this analysis which employed hypothesis testing in order to identify any significant differences.

		Environmental Concern					
		Unconcerned (n=13)	Indifferent (n=55)	Concerned (n=84)			
Age*	18-30	0.0%	31.3%	15.7%			
	31-50	89.8%	37.3%	42.5%			
	51-65	10.2%	25.2%	30.5%			
	65+	0.0%	6.3%	11.2%			
Gender*	Male	59.3%	74.6%	60.7%			
	Female	40.7%	25.4%	39.3%			
Driving Style*	Sporty	38.2%	26.5%	23.0%			
	Economic	24.2%	18.7%	20.9%			
	Safe	37.6%	54.8%	56.1%			
CO <sub>2</sub> per KM*	Mean	.1468	.1382	.1360			
	S.D	.3794	.0526	.0750			
KM per Trip	Mean	21.55	19.97	19.53			
	S.D	28.60	27.83	26.78			
Speeding Time*	Mean	246.23	254.92	219.99			
	S.D	256.07	368.65	347.63			

Table 13 Characteristics of project participants grouped by their stated level of environmental concern

\* Between group differences significant at the .000 level

Examining participants who stated they are unconcerned about the environment, this group tends to be populated by participants between the age of 30 and 50, with a somewhat even gender split whilst having a higher degree of individuals who either classify their driving style as sporty or economic. This finding might suggest that the desire of this group to drive in an economic manner is perhaps motivated by non-environmental issues, such as reducing vehicle operating costs. However, unconcerned participants also emit the highest quantity of  $CO_2$  emissions per kilometer, which may indicate that the participants who are included in this group and specified a sporty driving style are more than compensating for those who stated an economic driving style.

For participants who stated they are indifferent towards the environment, this group displays a somewhat even spread over the first three age categories whilst containing a significantly higher number of males compared to the other two groups. In terms of driving style, indifferent participants hold a similar distribution to that of concerned drivers across the three options. Additionally, this participant group also tends to speed for longer periods compared to unconcerned and concerned drivers.

The final participant group covers individuals who stated they are concerned about the environment. This group contains a higher proportion of older participants, with a relatively high percentage of drivers aged over 50. In terms of their driving style, concerned participants have a higher tendency to describe their driving as safe which is perhaps more linked to the age profile of this participant group than their environmental concern. The safe driving tendencies of this group are further supported by the time spent speeding, with concerned participants having the lowest average across all three groups on this metric.

What this analysis demonstrates is that a single question which asks individuals to express their level of environmental concern proves to be an effective method of partitioning drivers into unique groups which display significant differences on a number of related characteristics. The identification of questions which allow for an effective partition of individuals might prove valuable in an attempt to find a working method which can reduce complex market segmentation solutions (Anable, 2005) into a much smaller number of key questions or variables.

#### 4. Conclusions

An examination of the overall trend seen within the data would suggest that the provision of coaching could help to decrease emissions arising from driving. All experimental groups where some form of coaching was providing display lower emissions per kilometre values than the control group. An examination of the impact of introducing coaching to drivers was provided by Groups C and D, which received in car coaching and WEBFLEET, and WEBFLEET respectively. While both groups demonstrated decreased levels of emissions, it is not possible to ascertain whether the in car coaching significantly improved driver performance.

### Acknowledgements

The authors would like to thank CURE and TomTom for their help with conducting and collecting the data for this trial. This work was sponsored by the European Commissions PEACOX Project under the Seventh Framework Programme (FP7).

#### References

Anable, J., 2005. "Complacent Car Addicts" or "Aspiring Environmentalists"? Identifying travel behaviour segments using attitude theory. Transport Policy 12, 65–78.

- Barkenbus, J., Eco-driving: An overlooked climate change initiative. Energy Policy 38, 2010, 762-769.
- Beusen, B., Broekx, S., Denys, T., Beckx, C., Begraeuwe, B., Gijsbers, M., Scheepers, K., Govaerts, L., Torfs, R., Panis, L. Using on-board logging devices to study the longer-term impact of an eco-driving course. Transportation Research Part D. 14, 2009, 514-520.
- Boriboonsomsin, K., Barth, M., Vu, A., Evaluation of driving behaviour and attitude towards eco-driving: A Southern California Limited Case Study. 90th Annual Meeting of the Transportation Research Board, Washington, D.C, January, 2011.
- Delhomme, P., Cristea, M., Francoise, P., Self-reported frequency and perceived difficulty of adopting eco-friendly driving behaviour according to gender, age, and environmental concern. Transportation Research Part D. 20, 2013, 55-58.

- Martin, E., Boriboonsomisn, K., Chan, N., Williams, N., Shaheen, S., Barth, M. Dynamic ecodriving in Northern California: A Study of survey and vehicle operations data from an ecodriving feedback device. 92nd Annual Meeting of the Transportation Research Board, Washington, D.C, January, 2013.
- Martin, E., Chan, N., Shaheen, S. Understanding how ecodriving public education can result in reduced fuel use and greenhouse gas emissions. 91st Annual Meeting of the Transportation Research Board, Washington, D.C, January, 2012.
- Stillwater, T., Kurani, K. In-vehicle ecodriving interface: Theory, design and driver responses. 91st Annual Meeting of the Transportation Research Board, Washington, D.C, January, 2012.
- Rutty, M., Matthews, L., Andrey, J., Del Matto, T. Eco-driver training within the City of Calgary's municipal fleet: Monitoring the impact. Transportation Research Part D. 24, 2013, 44-51.
- Stromberg, H.K., Karlsson, M. Comparative effects of eco-driving initiatives aimed at urban bus drivers – Results from a field trial. Transportation Research Part D. 22, 2013, 28-33.