

Article



Green Driver: Travel Behaviors Revisited on Fuel Saving and Less Emission

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Abstract: Road transportation is the main energy consumer and major contributor of ever-increasing hazardous emissions. Transportation professionals have raised the idea of applying the green concept in various areas of transportation, including green highways, green vehicles and transit-oriented designs, to tackle the negative impact of road transportation. This research generated a new dimension called the green driver to remediate urgently the existing driving assessment models that have intensified emissions and energy consumption. In this regard, this study aimed to establish the green driver's behaviors related to fuel saving and emission reduction. The study has two phases. Phase one involves investigating the driving behaviors influencing fuel saving and emission reduction through a systematic literature review and content analysis, which identified twenty-one variables classified into four clusters. These clusters included the following: (i) FE_{f1} , which is driving style; (ii) FE_{f2} , which is driving behavior associated with vehicle transmission; (iii) FE_{f3}, which is driving behavior associated with road design and traffic rules; and (iv) FE_{f4}, which is driving behavior associated with vehicle operational characteristics. The second phase involves validating phase one findings by applying the Grounded Group Decision Making (GGDM) method. The results of GGDM have established seventeen green driving behaviors. The study conducted the Green Value (GV) analysis for each green behavior on fuel saving and emission reduction. The study found that aggressive driving (GV = 0.16) interferes with the association between fuel consumption, emission and driver's personalities. The research concludes that driver's personalities (including physical, psychological and psychosocial characteristics) have to be integrated for advanced in-vehicle driver assistance system and particularly, for green driving accreditation.

Keywords: driver travel behavior; green driver; driver assessment; eco driving; fuel consumption; tailpipe emission; road transportation

1. Introduction

The transportation sector is the main fuel consumer in the world and road transport is responsible for 85% of the average global energy [1]. Fuel consumption trends from the global transportation

perspective depend on drivers' travel behavior [2]. United Kingdom (UK) consumes 33% of global fuel energy [3], followed by Portugal 40% [3], United States (US) 28% [4], China 19% [5], India 16.9% [6] and Malaysia 25.5% [7]. Likewise, the transportation sector is in second place of hazardous gas emission in the world [4]. The vehicle emission trends have been increasing. In particular, CO₂ in UK is 25% [3], followed by Portugal 20% [3], US 28% [8], Japan 19% [9], China 23% [5], India 28.33% [10] and Malaysia 22.9% [11]. The rapid growth in global emissions was driven by emissions from the road sector (Figure 1). As passenger transport consumes more energy compared to freight transportation, most CO₂ emissions belongs to road transport (Figure 2). The increasing number of cars has intensified the amount of environmental pollution, especially carbon dioxide (CO₂), carbon monoxide (CO), particulate matter (PM) and etc. The elevated global CO₂ emission has been increased by about 2.2% from 2012 levels and reached 32.2 GtCO₂ in 2013 [5]. Nevertheless, the annual growth of CO₂ was reported slightly lower by 2.5% since 2000 [2]. Approximately, 80% of road transportation oil consumption is highly associated with 23% of CO₂ emissions in 2013 [2]. Road transport is responsible for 85% of the average global energy used in transport and aviation both grew by 25% [12].

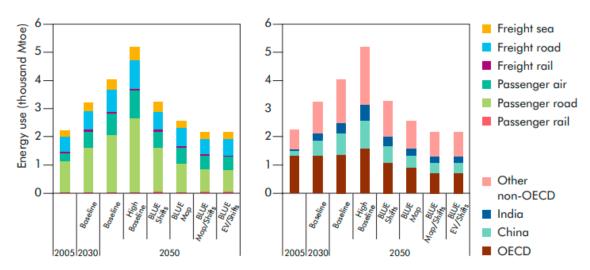


Figure 1. Global energy consumption by transportation modes (Passenger and Freight (2005–2050); (Source: International Energy Agency (IEA) [5]).

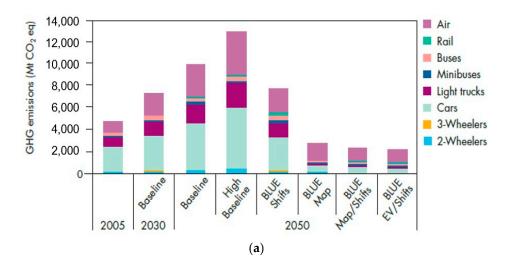


Figure 2. Cont.

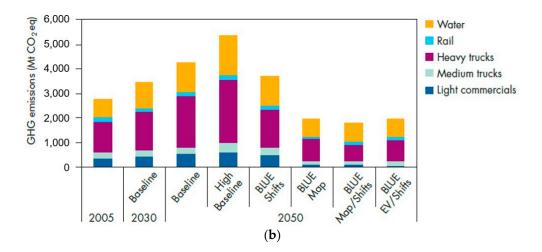


Figure 2. Global CO₂ emissions by transportation modes (**a**) Passenger; (**b**) Freight (2012–2040); (Source: International Energy Agency (IEA) [5]).

Travel behavior is the interaction between people and transport, which impacts to fuel consumption and CO_2 emission [13,14]. Recently, travel behavior researchers have created a new term called driver behavior, which considerably contributes to reducing fuel consumption and CO_2 emission. The driver behavior study involves three major components; included the driver, vehicle and environment [15]. The vehicle and environment are considered as physical factors, while driver has also psychological factors when measuring the driver's travel behaviors [15]. In fact, the driver is a complex factor that constantly changes, and it is unpredictable. Therefore, it is very hard to translate driver's behaviors accurately, due to the socio-demographic differences, preferences, beliefs, attitude, (including knowledge and self-awareness (i.e., insight) and upbringing issues. Besides, driver behavior is governed by emotions, mental activities and physical actions. The interaction between driver and environment or any physical surrounding may result in dramatic changes in the driver's behavior and his/her actions as well as feedback, which is called the behavioral dynamics of the driver. There has been extensive research on human behavior across physiology, psychiatry, health, neural development, personality and attitude, psychopathology and brain stimulus. In addition, various models have been developed and applied in numerous applications, such as artificial intelligence, risk assessment, decision making, system safety analysis, human performance and so on [16]. However, these approaches have not been yet integrated with travel behavior studies.

In particular, previous researchers have identified drawbacks that hindered driver's behaviors. Firstly, the repeatability and reliability of experiments is related to drivers' behaviors. Most of the time, it is impossible to obtain a reliable measurement of behavior, because the experimental work measures physical factors more than the psychological factors [17]. For example, in examining the effect of driver behavior on emission and fuel consumption, the types of engine, speed and acceleration have been considered as the main factors, while human-related factors have not been estimated. Thus, it is crucial to incorporate the psychological factors of a driver (i.e., personality traits) [18], attitudes and intentions [19] and risk-taking [20] for fuel saving and emission reduction studies. Moreover, driver cognitive factors, such as management of attention [21], visual functions [22] and psychomotor behavior [22], have to be explored for emission reduction and fuel saving studies. Hence, the integration of psychological and technical experiments first remains as the main gap in driver travel behavior studies. Secondly, the development criteria and attributes are critical to obtain good driving measurements [17]. In fact, there is no specific measurement method or technique to evaluate 'goodness' of a driver [17]. It can be described through various ways. For instance, the reduction of accidents and fatalities greatly reflect the safety practice of a driver and it could be taken as a measure of the 'goodness' of a driver [17]. In another example, driving at a constant speed will reduce fuel consumption compared to aggressive driving practices, which can also be taken as

a measure of goodness of a driver. Therefore, the existing in-vehicle driving assistance systems and simulators lack the incorporation of driver's behaviors on reducing energy consumption. Furthermore, the vehicle emission model stands independently. Limited exploration of drivers' competency in view of instantaneous decision making, planning, anticipation and driving goals as well as the scarcity of travel information makes it difficult to create fully accurate driving assessment models. Therefore, driver's travel behaviors could not be translated adequately, interdependently and rigorously through these models [23].

On the other hand, the green concept is no longer limited to the construction and material fields of research. It has been recently applied to the transportation sector and various approaches of the green concept have been studied, such as green highway design and construction [24], green public transportation [25,26], bio-fuel improvement [27], green vehicles (or, Eco-vehicles) [28] and vehicle safety features [29]. However, transportation, as the major contributor of ever-increasing CO₂ and greenhouse gas (GHG) emissions, has a significant role in green development agendas. In this regard, the transportation professionals and authorities have recently changed their approach to green transportation. The CO₂ and GHG emission reduction is the main goal in the growth of green transportation [30]. So, it is essential to innovate the techniques reduce CO₂ and GHG emissions [31,32]. Doppelt and Markowitz [33] state that "there does not appear to be substantial research on how specific behavioral changes can lead to measurable reductions in GHG emissions". Previous studies determined that the selection of modes of travel (i.e., transit, carpool, metro, bus and taxi) that emit less greenhouse gasses [34,35] but are not sufficiently successful to sustain driver's travel behavioral changes.

In this regard, the current study proposes a new angle to the green transpiration as; 'Green Driver' (see Figure 3), which may involve the driver's attributes, especially driver's psychological attributes in travel behavior studies. The green driver is defined as the driver's psychological attributes (including personalities, attitudes, cognitive and risks) contributing to fuel saving and emission reduction. Furthermore, the green driver concept may explore the driver's dynamics, actions and reflections with regard to their driving personality. The concept is governed by human factors and their interaction with surrounding (i.e., environment and vehicle).

According to above-mentioned issues and gaps, the current study aimed to establish the green driver behaviors for achieving fuel saving and emission reduction. To achieve the aim, the research has engaged two objectives; (i) to identify and establish the driver's green travel behaviors influencing fuel saving and emission reduction; and (ii) to estimate the Green Value (GV) impact of those green travel behaviors. Corresponding to these objectives, the research methodology was structured into two phases. In phase one, a comprehensive list of driver's travel behaviors incorporated to fuel consumption and tailpipe emission is first investigated through content analysis and systematic literature review methods. After this, the green travel behaviors out of all travel behaviors are identified and established using the Grounded Group Decision Making (GGDM) method. In the phase two, Green Value (GV) of these established green travel behaviors is analyzed, which indicates the green impact degree of each green behavior on fuel saving and emission reduction. The current study reports the requirements of developing the green driver behavior assessment index model and the developed model will be presented in future works.

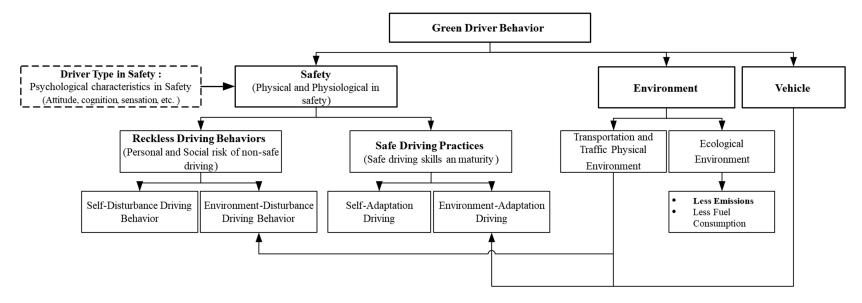


Figure 3. The Green driver conceptual framework.

The current study has drawn several scopes based on the environmental conditions, road design and rules, vehicle type, various circumstances and driver's characteristics. The green driver behaviors can be considered to be almost same in all nations. However, these behaviors might be adjusted according to climate and weather conditions [36]. In this regard, Yan et al. [37], Jung et al. [38] and Hamdar et al. [39] stated that dynamic nature of driving is affected by weather conditions. The abnormal weather and other environmental conditions factors substantially impact traffic flow dynamics [40] as well as traffic congestion and crashes [39,41], which may affect driver behavior (for example, resulting in speed reductions). Ibrahim and Hall [42] has discovered that free-flow speed decreases by 23.6–31 mph during heavy snow, 1.9 mph in light snow and 3.1–6.2 mph in heavy rain. Evans [43] found that foggy weather significantly impacts the behavior of drivers in following cars, with reductions in speed and acceleration rates. However, Hamdar et al. [39] stated that quantifying the correlation between driving behavior and weather conditions requires further studies on the micro-scale. In particular, there are very few studies on the association of weather conditions with fuel consumption and emission factors. Therefore, the present study has considered the weather conditions as a control variable. In addition, this research has shown that the legislative bodies of governments and local authorities of countries have imposed their own road driving rules, while there are minor differences. For example, the speed limit is almost same around the world, but the nations have adjusted it according to their weather conditions, road designs and drivers' modifiable factors (i.e., personalities, attitudes, cognitive and risk). Furthermore, this research has focused on petroleum-based vehicles since approximately 80% of current road fuels are derived from petroleum [44], while more than 75% of total CO_2 emission was emitted from gasoline and diesel consumption [5]. Moreover, the current research explores the green driving behaviors without focusing on the level of experience (i.e., beginner, experienced levels), while more experienced drivers may influence fuel saving more effectively than beginners. As the scope of this research is on vehicle and environment dimensions, the association between the green driver's behavior and safety will be reported in the next paper.

2. Identifying the Driver's Travel Behaviors Affecting Less Fuel Consumption and Tailpipe Emission

2.1. Systematic Literature Investigation

Phase one identifies the comprehensive list of travel behaviors incorporate tailpipe emission and fuel consumption. The research conducted a systematic review analysis, as it is replicable, scientific and transparent method of literature review [45]. This method can minimize bias on literature review as it provides comprehensive decisions, procedures and conclusion from researchers. The variables (i.e., travel behaviors) have been collected through a systematic literature review using specific keywords and combination of keywords (included, driver behavior, travel behavior, vehicle fuel consumption, vehicle tailpipe emission and green transportation) in the available references. Both cited and referenced articles were extracted from ScienceDirect, Google Scholar, Scopus, Taylor and Francis, Emerald, Sage as the available databases. In specific, most of the articles downloaded were retrieved from specific related journals; included, Transport reviews, Journal of Automobile Engineering, Transportation Research Part F: Traffic Psychology, Transportation Research Part D: Transport and Environment, Transport Policy, Journal of the Transportation Research Board, Transportation Science, Atmospheric Environment and Government reports, Inventories, Database and Conference papers.

The methodology for a systematic literature was divided into four (4) stages; identification, screening, eligibility and synthesizing (Figure 4). Identification stage involved the process of collecting articles using relevant keywords combination including, "vehicle fuel consumption", "vehicle tailpipe emission", "driving behavior", "driver behavior", "eco driving" and "green transportation". Total 100 numbers of articles were found in association with this study. However, only 67 numbers of articles were included and reviewed for this study which published between 1995 and 2017. According to Ariens et al.'s [46] and Lievense et al.'s [47] instructions, the identified

literatures have been screened in terms of the consistency and quality of the findings. Among all screened articles, there were only 83 articles that incorporated driver factors in relation with energy and emission while the rest of articles have purely focused on vehicle and environmental factors. Six (6) articles were backdated and non-retrieved; hence deducted from the list. Next stage was called as eligibility. Those 71 articles were clustered into three (3) categories; highly relevance, moderately relevance and less relevance to verify the eligibility of the documents correspond to the topic of this study. In eligibility stage four (4) articles were excluded since had poor relevancy to this study. The eligibility stage was completed by 6 September 2017 and in the synthesizing stage the final 67 articles have been scanned in-depth.

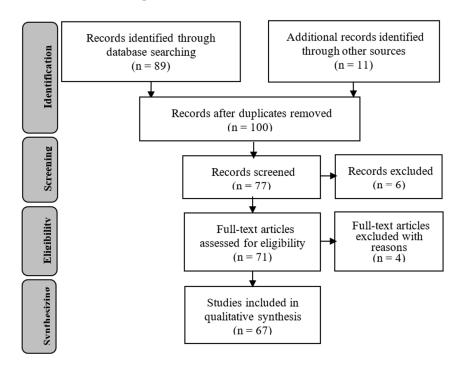


Figure 4. Systematic literature review flowchart on green driver behaviors.

Through content analysis method, all final 67 articles have been synthesized base on the codes (i.e., variables) extracted from each article. The result of synthesizing stage was tabulated in the form of content analysis table, which is presented in the following sections.

2.2. Literature Content Analysis and Synthesize

Table 1 presents collective literature on number of variables (i.e., driver behaviors) influence vehicle fuel consumption and tailpipe emission. The variables are classified into four (4) clusters (i.e., $FE_{f.}$) which are; (i) $FE_{f1.}$ Driving style; (ii) $FE_{f2.}$ Driving behavior associated with vehicle transmission; (iii) $FE_{f3.}$ Driving behavior associated with road design and traffic rules and (iv) $FE_{f4.}$ Driving behavior associated with Vehicle operational characteristics. The following explains each cluster and involved variables;

(i) FE_{f1} . Driving style; Previous literature has found that human behavior is a broad and uncertain subject as it is morally relative and differs from one person to another. There have been extensive studies conducted on human behavior, including physiology, psychiatry, health, neural development, personality and attitude, psychopathology and brain stimulus. Furthermore, various models have been developed and applied in numerous applications, such as artificial intelligence, risk assessment, decision making, system safety analysis, human performance, education etc. [48]. Most of the preferred developed models are based on the artificial neural network or genetic algorithms as it can generate and replicate substantial information in human research. Driving is also highly associated with human factors. Driving requires attention and multiple skills. Inattentive driving can possibly cause high energy (i.e., fuel) consumption and high emissions [43,49]. Driver behaviors are incorporated with human's mental and psycho-physical coordination, which reflects the ability of a driver to perceive driving situations and deal with it, for example at the intersections. Eventually, driving behavior can alter or reveal drivers' profile and personality after being subjected to the driving situations and road conditions. The behavior of drivers could be passive or aggressive, ignorant, alert, tolerant, competent, skillful and etc. According to De Vlieger [50], the amount of fuel consumption and tailpipe emission from aggressive driving increases by 20–40%, while calm driving results in a reduction of 5% compared to the normal driving. The existing driving behavior models capture drivers' tactical decisions in numerous traffic conditions. In the traffic engineering, there are two types of driving behavior models: macroscopic and microscopic models. These models are basically developed to compute and analyze traffic safety, traffic capacity and flow characteristics. Macroscopic models encompass the overall average traffic behavior in a system using partial differential equation (PDE) with functions for density, velocity and flow. Microscopic models measure the motion of individual vehicles in a system using the ordinary differential equation (ODE) with functions for position, velocity and acceleration. Macro-models are very general on a larger scale compared to the micro-models, which focus more on vehicle dynamics. However, massive computational work is required as the ODE computes results in each vehicle. Macro- and micro-models are further segregated into sub-models, such as Aw and Rascle (AR) model, Zhang model, gap acceptance model, car-following model, LightHill, Whitham and Richards (LWR) model and lane-changing model [26]. However, these models have limited exploration of drivers' competency in view of instantaneous decision making, planning, anticipation and driving goals. Moreover, scarcity of travel information and database has reduced the accuracy of these developed models. As a result, driving behavior cannot be translated adequately, interdependently and rigorously [23].

Recently, the concept of eco-driving has gained interest among researchers, which opposed the traditional driving style. Eco-driving reduces fuel consumption and emissions without reducing the average speed of vehicles [51,52]. Furthermore, it ensures a safer journey and increases social responsibilities [53]. Evans [43] recommended improving fuel economy by adjusting driver's behavior to follow specific instructions regarding improving fuel economy, for example, to minimize trip time; to avoid vigorous acceleration and deceleration. Eco-driving has potential in fuel saving [54]. The International Energy Agency [5] and previous studies [2,55–57] have calculated various amounts of fuel consumption savings from eco-driving and reported the short-term and medium-term impacts of eco-driving training. Just after eco-driving training, the average fuel efficiency will be improved by 15% [5,54,58], while, in the long-term, 5% to10% improvement will be expected [59,60].

(ii) FE_{f2}. Driving behavior associated with vehicle transmission; The researchers declared that the association between driving behavior and vehicular emission is not limited to driving characteristics (i.e., acceleration, deceleration, speed, idling, braking) [61]. There is evidence that speed and acceleration are the most preferred driving parameters used by previous studies to visualize energy consumption and emission in driving behavior studies. Journard et al. [62] reported that CO emissions decrease linearly with speed. He used vehicles with additional instruments to understand the correlation between emissions with acceleration and speed. For instance, 10 g/h of NO_X emissions were released at a constant vehicle speed (45 km/h) or zero acceleration. At negative acceleration (-5 km/h^2) , the amount of NO_x emission is estimated as 5 g/h, while this is approximately 18.5 g/h for positive acceleration (+5 km/h²) [62]. Moreover, previous research has stated that the posted speed can enhance fuel efficiency by 2–18%, if the drivers comply with lower speeds. At the peak engine torque, the fuel efficiency is lower. However, at higher RPM (Revolutions per minute) (low engine torque), acceleration is maximized and thus, fuel efficiency is higher [63–65]. Nevertheless, quick acceleration is still recommended by experts, although this should be a smooth transition. Enhancing fuel efficiency when driving requires drivers to anticipate the upcoming traffic or driving situations ahead in order

to optimize vehicle acceleration and braking [66]. Rolling resistance (tires > 25%) influenced fuel consumption by 3-5% [67].

(iii) FE_{f3} . Driving behavior associated with road design and traffic rules; Literature has shown that road and street design and traffic rules play significant roles in vehicle fuel consumption and emission. For example, Varhelyi [68] has investigated the effect of roundabout replacement with signalized intersection and vice versa. However, at an average junction that was rebuilt as roundabout, fuel consumption increased by 3%, CO emissions by an average of 4% and NO_x emissions by 6%. Thus, a large reduction in fuel consumption can be obtained by replacing roundabouts with signalized junction. However, it would still yield an equal amount if the roundabout was rebuilt with the signalized junction [69]. The effects of environmental factors (e.g., traffic volumes, traffic signs and controls towards the vehicle fuel and tailpipe emission) have also been studied. According to Al-Ghandour [70], the overall average of CO₂ emission and fuel consumption in the roundabout was reduced to 27% and 26%, respectively.

Green driving is associated with the driver's awareness of his/her driving context. The driving context refers to all features of travel behavior and traffic, such as traffic density, headway margin, traffic congestion, lane position, vehicle speed, aggressive driving, deceleration, acceleration, idling, etc. The driver's awareness refers to the driver's knowledge about their capabilities and limitations. Some studies have demonstrated that the driver's awareness of headway and traffic density helps to save the fuel and consequently, reduces emission of hazardous gases. Vaezipour et al. [71] stated that reducing headway by maximizing the speed, rapid acceleration or driving in the highest possible gear leads to a disruption in the traffic pattern, which may increase traffic density. Vaezipour et al. [71] suggested that drivers should stay in a fuel-efficient gear and use in-vehicle systems (e.g., intelligent speed adoption, congestion assistants, satellite navigation systems, Intelligent Transport Systems, etc.) to control the spacing margin between vehicles (i.e., headway). Nasir et al. [72] stated that traffic density affects vehicle average speed and the traffic flow rate. A green driver is an assertive driver, who is aware of different traffic conditions and monitors his/her reflection constructively to avoid accidents with other drivers. Therefore, green driver make headway without obstructing other drivers and maintain the safe distance to other vehicles without compromising traffic flow. Tang et al. [73] have investigated the impact of traffic density to multiple headways from the perspective of fuel consumption. They found that at a 20-m initial headway, a lower car density (20 nos) results in 13.4 L/100 km fuel consumption compared to a higher car density (80 nos), which consumes 13.1 L/100 km under the influence of ramps. At a 40-m initial headway, a lower car density (20 nos) results in 10.78 L/100 km fuel consumption compared to a higher car density (80 nos), which consumes 10.89 L/100 km under the influence of ramps. At an 80-m initial headway, a lower car density (20 nos) results in 10.28 L/100 km fuel consumption compared to a higher car density (80 nos), which consumes 10.46 L/100 km under the influence of ramps. [71]. Moreover, according to De Vlieger [50], the fuel consumption rate and tailpipe emission in rural and motorway driving modes were determined to be 30-40% lower compared to the urban area, which is normally 10 L/100 km.

Moreover, the emission factor is derived from specific vehicle driving cycles in different driving situations, such as urban, rural and motorway modes. The estimated emissions represent average results of travel scenarios, which should be applied at both the microscopic level and macro-scale (i.e., national/regional level) for emission inventories. The microscopic level models are suitable for green driving studies, as they tend to be more accurate in representing instantaneous operational driving actions. Although the macro-level models use the traffic data extracted from micro-level models, the microscopic level models might not be highly useful for green driving studies due to the limited trajectory data on aggregate driving characteristics. For example, the average speed model is derived from numerous speed levels in several trips. As a result, these are less meaningful on a microscopic level for green driving studies, because it does not represent specific driving characteristics [74].

(iv) FE_{f4} . Driving behavior associated with vehicle operational characteristics; The literature states that vehicle factors, such as types of engine, fuels, engine capacity, maintenance, air conditioned and tire pressure, have been shown to have a significant impact on tailpipe emission and fuel consumption. For instance, a conventional car consumes 12.6 L/100 km fuel with an average CO₂ emission of around 260 g/km, while the hybrid car consumes 8.23 L/100 km fuel with an average CO₂ emission of 180 g/km [75]. Vehicle model selection (e.g., eco-car) has the largest effect on fuel economy. Uneconomical fuel consumption was recorded as 800% for all passenger cars, 355% for fully electronic vehicles, 227% for fully electric and hybrid vehicles and 100% for all pickups. For example, the Toyota Prius (as a hybrid electric car) testing shows 62.6% fuel consumption of Yuanjian and 60% emission less than Yuanjian [76].

According to Table 1, the depth of citation by clusters shows that most previous studies have focused on driving behavior associated with vehicle operational characteristics 33.3%, driving behavior associated with vehicle transmission (32.5%) and driving behavior associated with road design and traffic rules (28.8%). Fewer travel behavior researches have been conducted from the perspective of driver factors, especially their driving style. That is a lack of research that shows linkages or relationships between humans and vehicles, humans and environment or humans and driving operations based on the collected papers and journals. The systematic literature review determined that the driving Style (5.4%) has not been sufficiently studied, which support this study's claim on lack of strong incorporation of driving behavior with vehicle emission and fuel consumption. Driving style here simply describes the types of drivers on the roads. Normally, aggressive drivers can be easily distinguished from other types of drivers (calm, sensible, normal, or eco).

To sum up, the road, traffic and driving behavior associated with vehicle operational characteristics should be measured and controlled. However, conducting experiments of human behavior is more complex and difficult to translate. Driver behavior deals with multiple personalities, emotions, attitudes, norms and beliefs. As human body and minds are inseparable, the communications between our emotional and rational systems can be witnessed throughout the decision-making process.

	FE _{f1.} Drivir	ng Style	FE _{f2.} W	Driving vith Veh	Behavi icle Trai	or Asso nsmissi	ociated on	FE _{f3}			vior Ass nd Traff			FE _{f4.}	Driving	; Behavi	or Asso	ociated w	vith Veh	nicle Oper	rational Characterist
Citations	FE _{f1.1} . Avoiding Aggressive Driving	FE _{f1.2.} Following Eco-Driving Rules	$\rm FE_{2.1}$ Avoiding Fast Acceleration/Deceleration of Vehicle	${ m FE}_{B.2.}$ Avoiding Excessive/Inappropriate Vehicle Speed	FE _{f2.3} . Caring of Inefficient Braking	FE $_{124}$. Caring of Inefficient Change of Gear	$FE_{2.5.}$ Avoiding Idling of Engine	$\mathrm{FE}_{\mathrm{fd},\mathrm{I}}$. Managing the Traveling Time and Traveling Distance	$\mathrm{FE}_{\mathrm{fd},2}$ Paying Full Attention and Obeying Traffic Signs and Signal	$FE_{f3.3}$ Paying Full Attention Traffic Calming Devices	$\mathrm{FE}_{\mathrm{f3,4}}$ Caring of Traffic Flow at Roadway/Street Intersections/Junctions	$\mathrm{FE}_{\mathrm{fd},\mathrm{S},\mathrm{M}}$ Monitoring and Adjusting to Traffic Volume	FE _{f3.6.} Minding Pedestrian Crossings	FE _{61.1.} Selecting the Most Suitable Fuel Type	$\mathrm{FE}_{\mathrm{fd},2}$ Aware and Monitoring the Capacity of Engine	FE _{14.3.} Aware and Monitoring Vehicles Horsepower	$\mathrm{FE}_{\mathrm{fd},4}$. Optimize Gear Shift and Transmission System Usage	$\mathrm{FE}_{\mathrm{fd},\mathrm{S}}$ Aware and Manage the Vehicle's Dead Weight	$\mathrm{FE}_{\mathrm{fd},6}.$ Regular Service and Check Tires Rolling Resistance	$\mathrm{FE}_{\mathrm{fd},7}$ Regular Service and Check Vehicle's Routing System	$\mathrm{FE}_{\mathrm{fd,R}}$ Monitoring and Cating of Vehicle's Maintenance
Robertson et al. [77] Descornet [78]				/					/		/							/	/		
Fischer [79] Kelly & Groblicki [80]				/				/			/	/		/	/	/		/	/		/
De Vlieger [50] Cicero-Fernández et al. [81]	/		/	/	/		/				/							/			/
Holmén & Niemeier [82] Wipke et al. [83]			/	/			/		/		/				/	/		/			
Höglund & Niittymäki [84] Ntziachristos & Samaras [85]			/	/			/	/		/	/	/									
Ericsson [86] Hydén & Várhelyi [87]			/	/		/		/	/	/	/			/	/	/		/			
van der Voort [88] Boulter [89]			/	/	/	/	/	/	/		/	/		/	/	/	/	/	/	/	
Ericsson [90] Rouphail et al. [91]			/	/		/		/	/		/				/	/					
Ahn et al. [92] Várhelyi [69]			/	/				/	/	/	/										
Cappiello et al. [93] Mandavilli et al. [94]			/	1			/		/	/	/		/		/		/	/			
Aerde et al. [95] Van Mierlo et al. [66]	/	/	/	/	/			/	/	/	/	/			/	/	/	/			/
Brundell-Freij & Ericsson [96] Weilenmann et al. [97]			/	/		/		/	/	/	/				 	 		/	/		
Daham et al. [98]			/	/	/			/	/	/	/								,		*

Table 1. Content analysis of driving behaviors influencing vehicle fuel consumption and tailpipe emission.

Table 1. Cont.

Depth of citation by cluster ercentage (%) Depth of citation by cluster	25				151					1	34			155							
Cluster	Driving s	style	Driv	ving beh vehic	le transn		with	Dri		n and a	issociate nd traffi		road]	Driving behavior associated with Vehicle operational characteristics						
Depth of citation	12	13	45	56	13	17	20	32	23	14	45	17	3	6	43	38	12	32	8	9	7
Srivatsa Srinivas and Gajanand [132]				/				/			/	/								/	
Zhao et al. [131]		,		,	,	/	/		,		'/	,									
Pampel et al. [130]		/	1	<i>'</i> /	/				/		<i>'</i> /	'/									
Tang et al. [71]			/	/							<i>'</i> /	'/						'		/	
Zhang et al. [129]			/	/		/	/	/	/		1	/						/		/	
Matsumoto & Peng [128]				/	/	',	/		',	/					/	/	/				
Al-Ghandour [70]				/		/	/	/	/	/		/		/			/	/			
Li et al. [127]		/			/	/		/				/		/					/	/	
Mensing et al. [125]		/	/	1		/	/	/	/							/		/	/	/	
Puan et al. [125]			/	1			/		',						/						
Matsumoto and Tsurudome [124]					/				/			/	/		/	/		/			
Li & Sun [123]					/	/	/	/					/								
Carrese et al. [122]	/	/			/		/				/	/								/	
Shafie-Pour & Tavakoli [121] Barić et al. [51]	/	/			,	/	,				/			/						/	/
Lee et al. [67] Chafia Davie & Tavialadi [121]							/	,			,			,							/
Vagg et al. [120]			,			/	,										/				,
Tulusan et al. [59]		/	/	/,	/	,									/	/	,	,			
Hari et al. [119]	/	/		/,	,	/		/							/	/,					
Iddo Riemersma [118]	,	,		1		,		,			/				/,	/,		/	/		
Laclair [117]				/,				/							/,	/,	/	1,			
Adriano et al. [116]		/		1	/			/,	/		/,		/		/,	/,	,	1,			
Prasad et al. [115]		,	,	,	,			,	,		,		,		/,	,		1,	,		
Wang et al. [114]				/				/													
Barth et al. [113]	/	/	/	/							/				/	/	/	/			
Sivak & Schoettle [65]	/	/		/			/				/	/			/	/	/	/			/
Fonseca González et al. [112]	/	/	/	/		/		/			/				/	/		/			
Kuo [111]				/				/										/		/	
Boriboonsomsin et al. [110]		/	/	/		/	/	/			/	/			/	/	/	/	/	/	
Shukla & Alam [109]			/	/			/	/			/			/	/	/					
Alessandrini et al. [108]	/		/					/			/					/	/	/			
Zlatkovic et al. [107]			,				,	/	/	,	,				/	/					
Li & Wang [106]			/	/			/	'		/	/							/		/	
Peng & Wang [105]				,				/	'		,				,	,		/		/	
Mulandi et al. [104]	,	/	'	<i>'</i> /				'	/	'	<i>'</i> /				<i>'</i> /	<i>'</i> /		/			
Strömberg et al. [103]	/	/	'/	/			/			/	'/	/			'/			'/			
Song et al. [76]				/			/	/			/				/	/		/			
Vermeulen [101] Greenwood et al. [102]		/	/			/	/		/	/					/	/	/	/			
Alessandrini et al. [75]	/	,	/		/	,	,		,	,		,					,	,			
Ericsson et al. [100]	,		,		,			/	/	/	/				/					/	
El-Shawarby et al. [99]	/		/	/						/	/	/				/	/				

3. Establishing the Driver's 'Green' Travel Behaviors Affecting Less Fuel Consumption and Tailpipe Emission

In second phase, an expert input study was conducted to determine the green value of each variable; and consequently, establish the green set of variables influencing fuel consumption and tailpipe emission in association with driving behavior.

The expert input study is such a group decision making procedure [133]. The group decision making procedure was conducted applying the grounded group decision making (GGDM) method. This method is a dynamic and heterogeneous decision-making process that incorporates selection of factors and consensus process upon factors selection. The GGDM is a useful method to overcome limitations in both technical and logistical aspect of the existing decision-making models. Logistically, GGDM can simplify time arrangement between decision makers and reduce cost of delay for decision making sessions. Technically, GGDM introduces limit and specific number of experts whom are well-versed and skillful in a particular area of study to remediate the difficulties in a discussed matter upon reaching consensus. Therefore, high number of experts are encouraged in GGDM process as it enhances sounder conclusion from different experts' points of view as well as voting powers in attaining consensus on the matter. In addition, the GGDM method allows experts to involve other experts who are well relevant and competent in that particular research area. In GGDM, the responsible decision maker called "GGDM researcher" which records the decision process and analyze the results [134–136]. If one of the proposed expert was duplicated or introduced again by the experts, the decision-making process will stop because GGDM method terminates any repetition of proposed experts in the loop in order to determine the most normalized final decision [136]. Lamit et al. [134] developed Equation (1), which $FW(a_i)$ extends to the final weight (FW) of issue number '*i*,' (a_i), of the discussion.

$$FW(a_i) = \left(\sum_{j=1}^n \left(\min\{WP_j, WPr_j\} \times SV_j\right)\right) \times a_i \tag{1}$$

where:

i, refers to criteria number (for *i* = 1, 2, 3, . . . , *m*).

a_i, refers to issue of discussion (i.e., Driver travel behavior variables).

 WP_{i} , refers to assigned weight by expert number 'j' in close group discussion for issue 'a_i'.

 WPr_j , refers to assigned weight by resource(s) relevant to the issue, whom introduced by expert number 'j' in close group discussion for issue ' a_i '.

 SV_{j} , refers to sessions value of the close group discussion sessions considered by the decision researcher for the set of issues ' a_i '.

Equation (2) indicates the consensus calculation of GGDM for issue ' a_i '. If the final consensus calculated more than 70%, the alternative is selected, and that criterion is approved.

$$FW(a_i)/FW(a_i)_{max} = \text{Consensus (\%)}$$
⁽²⁾

where, $FW(a_i)_{max}$, referred to maximum possible weight can be given to the issue ' a_i '.

A questionnaire was designed to be filled up by experts during group decision making sessions. The questionnaire form was designed using 5-point Likert scaling from 1 to 5 (from strongly disagree up to strongly agree). Two groups of experts consist of academician and practitioners were adopted in this study. The selection of expert is based on few criteria such as; background, specialization and experiences. The first group consists of lecturers, associate professor and PhD candidates across different disciplines in urban and transportation fields of research; including highway and transportation, traffic engineering, driver behavior, adaptive behavior, transportation planning and urban design and landscape architecture. The second group consists of practitioners across different

disciplines; included, traffic engineers, police officers and driving institute instructors. According to GGDM's purposive sampling, fourteen (14) participants have been involved; where eight (8) of them were academician and the remaining six (6) were practitioners. The GGDM method is able to overcome the logistical difficulties in arranging a specific time with involved decision makers and in saving the cost of managing the decision-making sessions. The GGDM method is also capable to manage the technical difficulties of experts involved in the decision making. The method also allows the appointment of voting power to each multiple participants of close group discussion (CGD). The result of the GGDM process is tabulated in Table 2 (academic experts) and Table 3 (practitioners).

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Table 2. Summary of GGDM data collection and analysis on variables influence vehicle fuel consumption and tailpipe emission in association with driving behavior from academic experts.

						v	alida	tion S	essio	n 1						V	alidat	ion S	essior	1 2		Validation Session 3								
		Par	ticipa	int 1	Pa	rticipa	ant 2	Par	ticipa	nt 3	Participant 4			Par	Participant 5		Par	ticipa	nt 6		Par	ticipa	nt 7	Part	licipa	nt 8		-	snsı	
Cluster.	Fuel Consumption & Tailpipe Emission Variables		WPr = c-WP5	c-WP	WP	r-WP	c-WP	WP	$WPr = WP1^*$	c-WP	WP	WPr = WP6	c-WP	SV	WP	WPr	c-WP	WP	WPr	c-WP SV WP	WPr = WP8	c-WP	WP	WPr	c-WP	SV	Cons. (%)	GGDM Consensu		
. ²⁰	FE _{f1.1.} Avoiding aggressive driving	5	5	5	5	-	5	5	-	5	-	5	5	2	5	-	5	5	-	5	3	5	4	4	4	-	4	1	98	Apprv
FE _{fl.} Driving style	FE _{f1.2.} Following eco-driving rules	5	3	3	4	-	4	4	-	4	-	5	5	2	3	-	3	5	-	5	3	4	4	4	3	-	3	1	79	Apprv
h c	$FE_{f2.1.}$ Avoiding fast Acceleration/Deceleration of vehicle	5	4	4	5	-	5	4	-	4	-	4	4	2	4	-	4	4	-	4	3	4	4	4	4	-	4	1	83	Apprv
wit] wit]	FE _{f2.2.} Avoiding Excessive/Inappropriate speed of vehicle	5	5	5	5	-	5	4	-	4	-	4	4	2	5	-	5	4	-	4	3	4	4	4	4	-	4	1	89	Apprv
E _{f2} . bel hicl	FE _{f2.3.} Caring of inefficient braking	4	3	3	3	-	3	4	-	4	-	4	4	2	3	-	3	4	-	4	3	5	4	4	4	-	4	1	71	Apprv
F ving socia ve ve	FE _{f2.4.} Caring of inefficient change of gear	4	3	3	4	-	4	3	-	3	-	4	4	2	3	-	3	4	-	4	3	4	4	4	3	-	3	1	70	Apprv
FE _{12.} Driving behavior associated with vehicle	FE _{f2.5.} Avoiding idling of engine	4	5	4	3	-	3	3	-	3	-	4	4	2	5	-	5	4	-	4	3	5	4	4	4	-	4	1	79	Apprv
	FE _{f3.1.} Managing the travelling time and travelling distance	5	5	5	3	-	3	1	-	1	-	4	4	2	5	-	5	4	-	4	3	4	4	4	4	-	4	1	76	Apprv
FE _{f3} . Driving avior associated with road design and traffic rules	$$\mathrm{FE}_{\mathrm{f3.2.}}$$ Paying full attention and obeying traffic signs and signals	2	3	2	4	-	4	3	-	3	-	4	4	2	3	-	3	4	-	4	3	5	3	3	3	-	3	1	66	Dis. Apprv
^{TE_{f3}. Dr avior as with r design}	FE _{f3.3.} Paying full attention to traffic calming devices	4	3	3	4	-	4	5	-	5	-	4	4	2	3	-	3	4	-	4	3	5	5	5	4	-	4	1	78	Apprv
FE _{f3.} behavior witi desi traff	FE _{f3.4.} Caring of traffic flow at roadway/street intersections/junctions	5	3	3	5	-	5	3	-	3	-	4	4	2	3	-	3	4	-	4	3	4	3	3	4	-	4	1	73	Apprv
	FE _{f3.5.} Monitoring and Adjusting to traffic volume	5	4	4	5	-	5	5	-	5	-	5	5	2	4	-	4	5	-	5	3	5	3	3	5	-	5	1	91	Apprv
	FE _{f3.6.} Minding pedestrian crossings	4	2	2	2	-	2	4	-	4	-	4	4	2	2	-	2	4	-	4	3	3	3	3	3	-	3	1	60	Dis. Apprv
	FE _{f4.1.} Selecting the most suitable type of fuel	5	4	4	5	-	5	1	-	1	-	4	4	2	4	-	4	4	-	4	3	4	4	4	3	-	3	1	74	Apprv
d stice	FE _{f4.2.} Aware and monitoring the capacity of engine	5	3	3	5	-	5	5	-	5	-	5	5	2	3	-	3	5	-	5	3	4	4	4	3	-	3	1	84	Apprv
FE _{f4} . Driving behavior associated with Vehicle erational characteristi	FE _{f4.3.} Aware of and monitoring vehicle's horsepower	5	3	3	5	-	5	5	-	5	-	4	4	2	3	-	3	4	-	4	3	4	4	4	5	-	5	1	80	Apprv
rivii ssoc ehic tara	FE _{f4.4.} Optimize gear shift and transmission system usage	4	3	3	4	-	4	5	-	5	-	4	4	2	3	-	3	4	-	4	3	5	4	4	5	-	5	1	78	Apprv
al ch V al ch V	FE _{f4.5.} Aware of and managing the vehicle's weight	4	3	3	5	-	5	5	-	5	-	5	5	2	3	-	3	5	-	5	3	4	4	4	4	-	4	1	85	Apprv
FE _{f4} . havio with tional	FE _{f4.6.} Regular service and check tires rolling resistance	3	2	2	3	-	3	5	-	5	-	4	4	2	2	-	2	4	-	4	3	4	4	4	3	-	3	1	66	Dis. Apprv
FE _{f4.} I behavior with ¹ operational c	FE _{f4.7.} Regular service and check vehicle's routing system	4	2	2	3	-	3	4	-	4	-	4	4	2	2	-	2	4	-	4	3	4	4	4	4	-	4	1		Dis. Apprv
б	FE _{f4.8.} Monitoring and caring of vehicle's maintenance	5	4	4	4	-	4	5	-	5	-	5	5	2	4	-	4	5	-	5	3	3	4	3	3	-	3	1	86	Apprv

Note. WP: Participant's Rate to the validation aspect, c-WP: conclusion of Participant's Rate to the validation aspect considered as the minimum of {*WPj*, *WPrj*}, WPr: Participant introduced resource Rate to the validation aspect, -: Participant did not provide value, SV: Close group discussion session Value considered by the GGDM researcher, Apprv: the validation aspect is approved based on GGDM Consensus rate of more than 70% agreement, Dis-Apprv: the validation aspect is not approved based on GGDM Consensus rate of not more than 70% agreement; WP1*: refers to the participant 1 in the non-academic expert Table 3 below.

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Table 3. Summary of GGDM data collection and analysis on variables influence vehicle fuel consumption and tailpipe emission in association with driving behavior from practitioner experts.

									Val	idation	Session	n 1										s	
		Participant 1			Pa	rticipar	nt 2	Pa	rticipar	nt 3	Pa	rticipar	nt 4	Pa	rticipar	nt 5	Participant 6				-	nsu	
Cluster	Fuel Consumption & Tailpipe Emission Variables		WPr = c-WP2	c-WP	WP	r-WP	c-WP	WP	WPr	c-WP	WP	WPr	c-WP	WP	WPr = c-WP6	c-WP	WP	WPr= WP8*	c-WP	sv	Cons. (%)	GGDM Conse	
fi. ing	FE _{f1.1.} Avoiding aggressive driving	4	4	4	4	-	4	4	-	4	3	-	3	4	5	4	5	-	5	2	80	Apprv	
FE _{fl.} Driving style	FE _{f1.2.} Following eco-driving rules	5	5	5	5	-	5	4	-	4	3	-	3	4	5	4	5	-	5	2	87	Apprv	
	FE _{f2.1.} Avoiding fast Acceleration/Deceleration of vehicle	5	5	5	5	-	5	4	-	4	3	-	3	4	5	4	5	-	5	2	87	Apprv	
FE ₇₂ . Driving behavior associated with vehicle transmission	FE _{f2.2.} Avoiding Excessive/ Inappropriate speed of vehicle	5	5	5	5	-	5	5	-	5	4	-	4	4	5	4	5	-	5	2	93	Apprv	
	FE _{f2.3.} Caring of inefficient braking	4	4	4	4	-	4	3	-	3	4	-	4	4	5	4	5	-	5	2	80	Apprv	
	FE _{f2.4.} Caring of inefficient change of gear	4	4	4	4	-	4	3	-	3	3	-	3	4	5	4	5	-	5	2	77	Apprv	
Dri ass tı	FE _{f2.5.} Avoiding idling of engine	4	4	4	4	-	4	4	-	4	3	-	3	4	5	4	5	-	5	2	80	Apprv	
	$\mathrm{FE}_{\mathrm{f3.1.}}$ Managing the travelling time and travelling distance	3	3	3	3	-	3	4	-	4	3	-	3	4	4	4	4	-	4	2	70	Apprv	
s sur	FE _{f3.2.} Paying full attention and obeying traffic signs and signals	3	3	3	3	-	3	3	-	3	4	-	4	3	4	3	4	-	4	2	67		
avic witi ign rule	FE _{f3.3.} Paying full attention to traffic calming devices	3	3	3	3	-	3	3	-	3	3	-	3	5	5	5	5	-	5	2	73	Apprv	
FE _R . Driving behaviour associated with road design and traffic rules	FE _{f3.4.} Caring of traffic flow at roadway/street intersections/junctions	3	3	3	3	-	3	3	-	3	3	-	3	3	5	3	5	-	5	2	67	Dis. Apprv	
F Driving associa road and tra	FE _{f3.5.} Monitoring and Adjusting to traffic volume	4	4	4	4	-	4	4	-	4	4	-	4	3	5	3	5	-	5	2	80	Apprv	
Д	FE _{f3.6.} Minding pedestrian crossing	2	2	2	2	-	2	3	-	3	4	-	4	3	4	3	4	-	4	2	60	Dis. Apprv	
	$FE_{f4.1}$. Selecting the most suitable type of fuel	4	4	4	4	-	4	2	-	2	3	-	3	4	5	4	5	-	5	2	73	Apprv	
r E	FE _{f4.2.} Aware and monitoring the capacity of engine	4	4	4	4	-	4	3	-	3	3	-	3	4	5	4	5	-	5	2	77	Apprv	
viou iion iics	FE _{f4.3.} Aware of and monitoring vehicle's horsepower	4	4	4	4	-	4	3	-	3	3	-	3	4	5	4	5	-	5	2	77	Apprv	
f4. ehav ed v berat erist	FE _{f4.4.} Optimize gear shift and transmission system usage	4	4	4	4	-	4	3	-	3	3	-	3	4	5	4	5	-	5	2	77	Apprv	
FE ₁₄ . Driving behaviour associated with Vehicle operational characteristics	FE _{f4.5.} Aware of and managing the vehicle's weight	4	4	4	4	-	4	3	-	3	3	-	3	4	5	4	5	-	5	2	77	Apprv	
brivi assc ehicl cha	FE _{f4.6.} Regular service and check tires rolling resistance	3	3	3	3	-	3	3	-	3	3	-	3	4	4	4	4	-	4	2	67	Dis. Apprv	
Ω ≯	$\mathrm{FE}_{\mathrm{f4.7.}}$ Regular service and check the vehicle's routing system	4	4	4	4	-	4	3	-	3	3	-	3	4	4	4	4	-	4	2	73	Apprv	
	FE _{f4.8.} Monitoring and caring of vehicle's maintenance	4	4	4	4	-	4	3	-	3	3	-	3	4	5	4	5	-	5	2	77	Apprv	

Note: Note: WP: Participant's Rate to the validation aspect, c-WP: conclusion of Participant's Rate to the validation aspect considered as the minimum of {*WPj*, *WPrj*}, WPr: Participant introduced resource Rate to the validation aspect, -: Participant did not provide value, SV: Close group discussion session Value considered by the GGDM researcher, Apprv: the validation aspect is approved based on GGDM Consensus rate of more than 70% agreement, Dis-Apprv: the validation aspect is not approved based on GGDM Consensus rate of not more than 70% agreement; WP8*: refers to the participant 8 in the academic expert Table 2 above.

3.1. Expert Input Analysis and Results

For an academic group of experts, there are four (4) validation sessions required upon reaching consensus. Validation sessions were differentiated according to the experience of involved experts in each session. The first validation session was conducted among academicians with less than 10 years of experience. The second validation session was conducted among academicians with more than 10 and 15 years of experience. The third validation session was conducted among researchers without practical working experience. While, the final validation session was conducted among traffic engineers, police officers and driving instructors with less than 10 years of experience.

Tables 2 and 3 show the summary of GGDM data collection and analysis results of experts' judgments to each variable significantly influencing vehicle fuel consumption and tailpipe emission in association with driving behavior. The following presents an example of GGDM method application and calculation process on the variable $FE_{f1.1}$. Avoiding aggressive driving.

Example (see Table 3):

FE_{f1.1.} Avoiding aggressive driving:

• Step 1—Calculate the score $\sum (C_{wp} \times sv)$; C_{wp} is the minimum score,

 $FW(a_i) = ((C_{wp1} \times sv) + (C_{wp2} \times sv) + (C_{wp3} \times sv) + \dots + (C_{wpn} \times sv)) = ((4 \times 2) + (4 \times 2) + (4 \times 2) + (3 \times 2) + (4 \times 2) + (5 \times 2)) = 48$

• **Step 2**—Calculate $FW(a_i)_{max}$, Max = 5

FW $(a_i)_{max} = (5 \times \text{sv} \times \text{No. of participants}) = (5 \times 2 \times 6) = 60$

• **Step 3**—Calculate the consensus rate (%)

Consensus (%) = $FW(a_i)/FW(a_i)_{max} = (48/60) \times 100 = 80\% > 70\%$, thus 'approved'

Results are tabulated independently based on groups of experts. The initial results in Table 2 shows; all variables have been approved as green variables exempt two (2) variables which are, $FE_{f3.2.}$ Paying full attention and obeying traffic signs and signals (66%) and $FE_{f3.6.}$ Minding pedestrian crossings (60%) were not approved since the consensus percentages were less than 70% saturation threshold. Meanwhile, in Table 3 shows three (3) unapproved variables which are; $FE_{f3.4.}$ Caring of traffic flow at roadway/street intersections/junctions (67%), $FE_{f3.6.}$ Minding pedestrian crossings (60%) and $FE_{f4.6.}$ Regular service and check tires rolling resistance (67%) for the same uncovered 70% saturation threshold. So, these variables have been dropped from the initial list. The next section measures the GV of these established green variables using the results of depth of citation (i.e., frequency of citation) reported in the contents analysis Table 1.

3.2. Green Value (GV) Analysis of Driver's Green Travel Behaviors

The green driver is defined as a driver who makes green-wise decisions in traveling focusing on safety, vehicle and environment, merges his/her psychological and psychosocial attributes and personality traits (i.e., preferences, beliefs, attitude, intention, cognition, attention and psychomotor behavior) with his/her technical knowledge in respect of reducing fuel consumption and emission. Accordingly, the Green Value (GV) assigns the green score which is drawn from evaluation and assessment of the driver's behaviors, actions and reflections across these green characteristics. To analyze the GV of each variable, this section integrates the results output from the content analysis Table 1 (i.e., depth of citation) and expert input results (i.e., GGDM analysis results). The purpose of this section is to establish the green impact of those established variables output from GGDM analysis procedure. This section indicates some variables will be eliminated instantly, although the depth of citation is higher than the GGDM consensus rate. The depth of citation is determined using the citation frequency divided by total number of cited (i.e., output from 4th stage of the systematic literature review process) articles. Meanwhile, GGDM results from two groups of academics and practitioners were summed and divided into 2, to obtain the average values. Then, both values are multiplied as shown in the Equation (3) below.

Green Value
$$(GV_{a_i}) = DC_{a_i} \times GC_{a_i} = \frac{\sum_{k=1}^{n} CR_{a_i}}{\sum_{k=1}^{n} R} \times GC_{a_i}$$
 (3)

where:

i, refers to criteria number (for: 1, 2, 3, ..., m).

k; refers to article number (for k = 1, 2, 3, ..., n).

DC_{ai}, refers to Depth of Citation of factor 'ai' (extracted from Table 1: Content Analysis).

GC_{ai}, refers to GGDM Consensus of factor 'ai' (extracted from Tables 2 and 3: GGDM result).

C, number of referenced articles involved in content analysis Table (extracted from Table 1: Content Analysis).

CR, number of articles have been cited for the factor 'a_i' (extracted from Table 1: Content Analysis).

The following presents an example of Green Value (GV) analysis and calculation of Equation (3) on the variable $FE_{f1.1}$. Avoiding aggressive driving (see Table 4).

Table 4. Data extracted from Tables 1–3 for the criterion 'Avoiding aggressive driving'.

Content Analysis (Table 1)	Expert Inputs (Tables 2 and 3)
Depth of citation = 12	GGDM consensus rate (academician) = 0.98
Total cited articles = 67	GGDM consensus rate (practitioners) = 0.80
Percentage depth of citation = $12/67 = 0.18$	Average GGDM consensus rate = $(0.98 + 0.8)/2 = 0.89$

Example:

FE_{f1.1.} Avoiding aggressive driving:

Refereeing to Table 4 data, thus;

So, Green Value (GV) $FE_{f1.1.}$ = Average depth of citation $FE_{f1.1.}$ × Average GGDM consensus rate $FE_{f1.1.}$ = 0.18 × 0.89 = 0.16.

Table 5 shows the GV analysis of the green variables. The aim of this section is to indicate the green value of each driver behavior to tailpipe emission ad fuel consumption. According to Table 3, four (4) variables were not applicable in GV analysis with respect to results of previous section on low consensus rate of GGDM calculation (included, $FE_{f3.4.}$ Caring of traffic flow at roadway/street intersections/junctions (67%), $FE_{f3.6.}$ Minding pedestrian crossing 60%), $FE_{f4.6.}$ Regular service and check tires rolling resistance (67%) and $FE_{f4.7.}$ Regular service and check the vehicle's routing system (69%).

		Depth o	of Citation		GGDM Result		
Cluster.	Variables –	Frequency	Percentage (%)	Academic	Practitioners	Average	— Green Value (GV)
FE _{f1} Driving style	FE _{f1.1.} Avoiding aggressive driving	12	0.18	0.98	0.80	0.89	0.16
TE _{fl.} Driving style	FE _{f1.2.} Following eco-driving rules	13	0.19	0.79	0.87	0.83	0.16
	FE _{f2.1.} Avoiding fast Acceleration / Deceleration of vehicle	45	0.67	0.83	0.87	0.85	0.57
	FE _{f2.2.} Avoiding Excessive/ Inappropriate speed of vehicle	56	0.84	0.89	0.93	0.91	0.76
FE _{f2.} Driving behavior associated	FE _{f2.3.} Caring of inefficient braking	13	0.19	0.71	0.80	0.76	0.15
with vehicle transmission	FE _{f2.4.} Caring of inefficient change of gear	17	0.25	0.70	0.77	0.74	0.19
	FE _{f2.5.} Avoiding idling of engine	20	0.30	0.79	0.80	0.79	0.24
	FE _{f3.1.} Managing the travelling time and travelling distance	32	0.48	0.76	0.70	0.73	0.35
	FE _{f3.2.} Paying full attention and obeying traffic signs and signals	23	0.34	0.66	0.67	0.67	* Not applicable
FE _{f3.} Driving behavior associated	FE _{f3.3.} Paying full attention to traffic calming devices	14	0.21	0.78	0.73	0.75	0.16
with road design and traffic rules	FE _{f3.4.} Caring of roadway and street design	45	0.67	0.73	0.67	0.70	0.47
	FE _{f3.5.} Monitoring and Adjusting to traffic volume	17	0.25	0.91	0.80	0.86	0.22
	FE _{f3.6.} Managing towards pedestrian crossings	3	0.04	0.60	0.60	0.60	* Not applicable
	FE _{f4.1} , Selecting the most suitable type of fuel	6	0.09	0.74	0.73	0.73	0.07
	FE _{f4.2.} Aware and monitoring the capacity of engine	43	0.64	0.84	0.77	0.80	0.51
FE _{f4.} Driving behavior	FE _{f4.3.} Aware of and monitoring vehicle's horsepower	38	0.57	0.80	0.77	0.79	0.45
associated with Vehicle	FE _{f4.4.} Managing and monitoring the vehicle's transmission	12	0.18	0.78	0.77	0.77	0.14
operational characteristics	FE _{f4.5.} Aware of and managing the vehicle's weight	32	0.48	0.85	0.77	0.81	0.39
operational characteristics	FE _{f4.6.} Aware of and managing the vehicle's tire rolling resistance	8	0.12	0.66	0.67	0.67	* Not applicable
	FE _{f4.7.} Managing and controlling the vehicle's routing system	9	0.13	0.65	0.73	0.69	* Not applicable
	FE _{f4.8.} Monitoring and caring of vehicle's maintenance	7	0.10	0.86	0.77	0.82	0.09

Table 5. Green Value (GV) analysis of green travel behavior variable	iables influencing less fuel consumption a	and tailpipe emission.
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Note: * Not applicable: The variable was non-approved in GGDM process and cannot be involved in GV analysis.

Also, Green Value (GV) is folded into three (3) clusters of green impact degree as; (i) ≤ 0.3 ; (ii) 0.3–0.5; and (iii) ≥ 0.5 .

- (i) if ≤ 0.3 ; has 'weak' contribution to the vehicle fuel and emission measurements.
- (ii) if 0.3–0.5; has 'moderate' contribution to the vehicle fuel and emission measurements.
- (iii) if ≥ 0.5 ; has 'high' contribution to the vehicle fuel and emission measurements.

According to Figure 5, driving speed (FE_{f2.2.}) (0.76) and vehicle acceleration (FE_{f2.1.}) (0.57) are categorized into the third cluster (i.e., high contribution), while vehicles associated with Vehicle operational characteristics included, capacity of engine (FE_{f4.2.}) (0.51), vehicle horsepower (FE_{f4.3.}) (0.45), vehicle weight (FE_{f4.5.}) (0.39), roadway and street intersection design (FE_{f3.4.}) (0.47) and traveling time and distance (FE_{f3.1.}) (0.35) are categorized in the second cluster (i.e., moderate contribution). Last but not least, the remaining variables which are less than or equals to 0.3 are categorized under first cluster (i.e., minimum contribution). In particular, minimum focus on aggressive driving (FE_{f1.1.}) (0.16) has hindered the association between fuel consumption, emission and human behavior. Hence, human factors remain unaccountable and non-measurable yet in green driver behavior studies. Scarcity of literature has evidenced that appropriate methods are intensely required to measure and translate drivers' aggressiveness or personality towards fuel consumption and emission while driving.

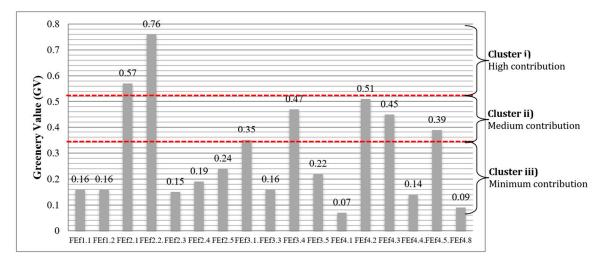


Figure 5. Green Value (GV) analysis results of driver's green travel behavior variables influencing less fuel consumption and tailpipe emission. The three (3) clusters of green impact degree are shown.

4. Discussions

The fuel consumption can be considerably reduced by short-term and long-term training [137], however, Johansson et al. [138] expressed that "both training and motivation are needed for drivers to achieve long-term fuel savings".

According to previous studies, drivers' travel behaviors can be clustered to the following: (1) behavior in response to values and beliefs; (2) behavior in response to skills and capabilities; and (3) behavior in response to car environmental quality. In respect to each type of behavior, drivers may make different decisions on fuel economy improvement strategies, including strategic, tactical and operational decisions;

• Strategic decisions can reduce the environmental impact of travel [66]. For example, maintaining optimal tire pressure and the emission control system are the major driver's strategic decisions, which can reduce 40% of vehicle emissions [139]. According to Sivak and Schoettle [66], in strategic decisions, the vehicle type has the greatest impact on (38%) the fuel economy. Congestion in tactical dimension and aggressive driving in operational dimension has more influence than other factors (40% and 30%, respectively).

- Tactical decisions aids to minimize the negative environmental impact of travel. Some tactical decisions include choosing smaller-engine vehicles [8], optimum choice of route [66], shortening the length/time of travel and choices of vehicle loading. According to EPA [8], each extra 45 kg of load causes a 2% increase in fuel consumption. The optimum choice of route contributes 15–40% increment in fuel economy [66]. This performance will be improved by 23% using a fuel-optimized navigation system [92].
- As operational decisions contribute to fuel consumption and emission reduction by 5–30% [110,140]. Training drivers to have more gradual and smooth driving style is an operational decision that helps to reduce fuel consumption in contrast with aggressive driving, hard acceleration and excessive speed [141].

In this research, twenty-one (21) driver travel behaviors were identified. All green travel behaviors (i.e., variables) were identified through a systematic literature review and were weighted based on the depth of citation (i.e., frequency of citation) and GGDM method's consensus rate. Four (4) variables were dropped according to GGDM consensus condition (including, $FE_{f3,2}$. Paying full attention and obeying traffic signs and signals, FE_{f3.6}. Managing towards pedestrian crossings, FE_{f4.6}. Aware of and managing the vehicle's tire rolling resistance and FE_{f4.7}. Managing and controlling the vehicle's routing system) and the rest Seventeen (17) variables have been established as the green travel behaviors. The green impact analysis was incorporated to these variables. The green impact analysis indicated three (3) clusters of variables; (i) Green Value (GV) is less than or equals to 0.3; (ii) GV is in between 0.31 and 0.5; and (iii) GV is greater than or equals to 0.51. The green impact analysis resulted these variables have been classified in the Cluster i, as moderate to highly significant influencing on less fuel consumption and less emission; included, avoiding excessive/ inappropriate speed of vehicle (0.76), avoiding fast acceleration / deceleration of vehicle (0.57), aware and monitoring the capacity of engine (0.51), caring of roadway and street design (0.47), aware of and monitoring vehicle's horsepower (0.45), aware of and managing the vehicle's weight (0.39) and managing the travelling time and travelling distance (0.35). Although the rest of variables can contribute to less consumption and emission, their impact would be minor.

For the speed variable, vehicle emission and fuel consumption are more sensitive to non-constant cruise-speed levels. Therefore, vehicle emission and fuel consumption rates will be considerably increased as a driver drives. However, converting the driver's speed behavior to accurate amount of fuel consumed based on acceleration is a big challenge. The studies show that aggressive driving affects speed and therefore, will have an impact on fuel consumption and emission. Ogle [142] states "although aggression is not a behavioral consideration in fuel and emissions studies, similar definitions based on appropriate acceleration rates could possibly be used as surrogate measures for speeding, following too closely, improper lane change and perhaps other aggressive driving behaviors". The research on driver speed behavior and acceleration have mostly focused to non-usual vehicles (i.e., hybrid cars, electric-cars, eco-cars, etc.).

For the acceleration variable, De Vlieger [50] identifies three types of driving behaviors associated with fuel consumption based on acceleration: calm driving (1–1.45 mph/s), normal driving (1.45–1.90 mph/s) and aggressive driving (1.90–2.45 mph/s). Nam et al. [143] and Zorrofi et al. [144] stated that aggressive driving behavior and higher fuel emissions are strongly correlated. Aggressive driving wastes gas mileage by 33% at highway speeds and by 5% around town [99]. Rakha and Ding [95] stated the constant deceleration rates (-0.25 to -1.50 m/s²) at increments of -0.25 m/s² reduced the amount of fuel consumption by about 0.09 L/km.

For the road and street variables, Ericsson [86] stated that the street type significantly impacts to the fuel emissions and the driver's aggressive acceleration behavior comes second. Hassan [145] stated "besides driver's error, road characteristics play a major role in collision occurrence." Furthermore, Austroads' reports [146] expressed that "the relationship between vehicle speed, curve radius, pavement super-elevation, friction between tire and road surface and gravity is well documented" if there is a comprehensive road design guidance. Most of the research on road geometric design focuses

on four areas: vehicle stability, operating speed, driver workload and alignment indices [147,148]. However, there are some factors, such as road slope, pavement, friction, weather conditions, level of skill, accident circumstances and so on, can be studied in future, as they would give a broader knowledge of how driver behavior is affected by road geometry and design features. Moreover, previous research has focused more on safety, especially as several studies have indicated that safety can be affected by road design. For example, Theeuwes et al. [149] expressed that the association between driver behavior and road design is a gap in previous research as it can reduce the accident severity and frequency. However, the triangulation between driver behavior, road design and energy saving/emission reduction has room to be analyzed and investigated.

For the vehicle horsepower variable, Brundell-Freji and Ericsson [96] state that the "power-to-weight ratio had a fairly large impact on overall driving style," which was supported by Andre et al. [150]. Berry [151] and Zhang et al. [129] stated that vehicle design changes (such as higher number of gears and lower power-to-weight ratios) can increase the efficiency. Berry [151] expressed that the power-to-weight ratio in moderate-performance vehicles is not an effective factor in aggressive driving studies, although the triangulation with energy saving and emission reduction was not yet studied.

For the capacity of engine variable, reviewing the literature demonstrates that no triangulation between the driver behavior, capacity of engine, vehicle emission and fuel consumption have been conducted. Therefore, studying such triangulation analysis would be essential for transportation scholars and practitioners.

For the vehicle weight variable, there are very general and few studies that have examined the correlation between driver behavior, vehicle weight and vehicle emission and fuel consumption. For example, Wu et al. [152] suggest generally avoiding unnecessary weight and removing roof racks.

For the traveling distance and time variable, literature has indicated that vehicular emissions–distance models can also be used to represent the real driving behavior. However, such models are sophisticated and complex, require massive amount of data and can be particularly used in a microscopic level studies. Technically, such models show that the standard deviation of accelerator pedal/throttle is lower and thus, it verifies the correlation between emissions and driving style [75]. In comparison to the earlier models, these models are formulated based on the acceleration and speed of vehicles under different driving operations, which are idle, acceleration, deceleration and steady-state cruising speed. Even so, there are still gaps in the vehicular emission model development, especially in variation of vehicles and driving characteristics [74,153].

5. Conclusions

The behavior assessment of drivers is vital to improve the existing human travel behaviors. This study has investigated numerous driver behaviors that influence vehicle fuel consumption and tailpipe emission in association with driving behavior. According to the analysis, driver's modifiable factors (i.e., personalities, attitudes, cognitive and risks) are seldom underestimated in consideration of fuel consumption and emission in driver travel behavior studies. Previous researchers were keen to investigate the environmental and vehicle effects on fuel and emission as these components have become more translatable and measurable compared to human behavior. This study declares that driver behavior is rather complex and is governed by psychological, psychosocial and cognitive attributes. Indeed, driver behavior can be highly associated with how well his/her brain makes decisions and translates the information through motor skills in driving activities. Thus, we highly recommend further investigation into driver behavior in fuel consumption and emission studies, with this study having explored and established a set of seventeen behaviors. The outcome of this research has constructed the firm platform for developing the green driver behavior index model, which can assess the green of any individual driver in fuel-saving and low-emission style of driving. To develop this index model, the established list of driver behaviors will be applied by researchers to carry out experimental works on measuring green driver behaviors. This index model will make the

driver's behavior more measurable in the future. Furthermore, the index model will able to explore and understand the driver's mindset and decision-making patterns for green driving.

Transportation planners, urban designers and planners, consultants, authority may use this comprehensive list of seventeen green driver behaviors as a checklist for green accreditation, assessment and labeling the driver. Furthermore, the research states that weighted value (i.e., coefficient) of each behavior can be applied in the driver assessment indexing formula.

This research reports the first phase of a big project. This phase has shown that the development of continuous and advanced monitoring devices to assess driving behavior is highly treasured and thus, the green driver concept would be more measurable in future travel behavior studies. Particularly, psychological and personality traits could be incorporated into any established green travel behaviors and studied thoroughly. In the next phase, the established list of the green behaviors can be incorporated into the development of the green driver index assessment tool. In future works, correlation analysis between green driver behaviors and vehicle emission fuel consumption can be studied. Indeed, these established green travel behaviors can be used as a checklist to evaluate the driver's performance with respect to green driving. In addition, this checklist can be upgraded to a stand-alone and/or web-based decision support system for assessing, monitoring, recording and publishing reports on a driver's performance in green driving accreditation.

The findings of this study can be generalized to all nations, although according to the explained scopes of research in the introduction, there are some minor difference in terms of policies, culture, climate and so on of the nations. Hence, the list of green behaviors can be adjusted and expanded according to governmental transportation planning rules and policies as well as other related issues.

The Green Value (GV) index includes 2 parts: DC_{a_i} and GC_{a_i} . The DC_{a_i} is the coefficient for GC_{a_i} and thus, DC_{a_i} may have a minor impact on the final GV result. As the systematic literature review method has been applied to obtain the depth of the citation for each criterion, any changes to the systematic procedure may affect the DC_{a_i} , although the impact of this on the final GV of that criterion should be very minor. The changes to this systematic procedure would be due to the sources available for literature collection, updated recent articles, increased citation of the selected articles and so on. Moreover, in this research, the GV was the measure of the association of green driver behaviors with fuel consumption and emissions through an expert input study so it was estimated qualitatively. As previously mentioned, we suggest that future works should measure this quantitatively as well.

At the end, this paper scoped to environment and vehicles dimension of the green driver concept, while the association between green driver behavior and safety will be investigated in future works.

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