

Providing a foundation for road transport energy demand analysis: A vehicle parc model for South Africa

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Abstract

It is key for national economic planning to build the tools to forecast energy demand from major sectors like transport in a credible way. As a starting point, this requires building a sufficiently detailed 'bottom-up' picture of technologies and their activity levels in the recent past. A vehicle parc model was developed for South Africa to feed transport demand and data on the fleet into a national energy systems model, the South African TIMES model, which is a least-cost optimisation model of the TIMES/MARKAL family. Detailed assumptions were developed for 24 vehicle typologies that included the vintage profile, annual mileage and its relationship with age, fuel economy and its improvement over time, and occupancy and load factor. Combining these assumptions, the model was successfully calibrated over 2000–2014 with the national registration database, national fuel sales statistics and, on the freight side, with estimates of the demand for ton.km published by the University of Stellenbosch's Department of Logistics (2014 only). A demand for passenger.km was also calculated, which agreed well with national transport surveys. A

range of detailed indicators were produced for the vehicle typologies and some interesting trends observed, including the steady dieselisation of the light vehicle fleet over the study period and the stagnation of passenger car fuel economy, despite legislation in the European Union. The present study believes that this updated data-rich picture of the road transport vehicle parc will support other studies and national policy and planning initiatives.

Keywords: freight demand, fuel economy, mode share, greenhouse gas emissions, vehicle fleets, modelling

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1. Introduction

The national energy balance for 2013 indicated that the transport sector accounted for an estimated 30% of South Africa's total final consumption (Department of Energy (DoE), 2013). The sector has remained almost completely dependent on liquid fuels, which accounted for 98% of demand in the sector in 2013 and 83% of all liquid fuels used in the economy (DoE, 2013).

Large investments and long lead times are involved in meeting the demand for liquid fuels and transporting liquid fuels from point of supply to the point of demand. In addition, the choice of primary energy and the transformation process can have substantial impacts on society and the environment. Investment decisions must, therefore, be informed by planning processes such as a national integrated energy plan. The first step in the planning process is to build an understanding of the current demand for mobility of passengers and freight in the economy and the drivers of mobility in the transport sector and develop credible scenarios of how these will evolve over time. Furthermore, the need for mobility is not something that can be directly measured or observed and, therefore, requires estimation based on a number of observable variables such as how many people are driving private vehicles, the demand for the movement of goods in the current economic environment, and how many vehicles are on the road network (Merven, et al., 2012).

The Energy Research Centre at the University of Cape Town and the South African National Energy Development Institute used the South African TIMES Model (SATIM) in 2012, which is a least-cost optimisation model of the TIMES/MARKAL family, to assess the demand for energy from transport to 2050 (Merven, et al., 2012). The outputs of that study were widely applied (Department of Environmental Affairs, 2014; DoE, 2012; DoE, 2016; Gajjar & Mondol, 2015). A series of papers, of which this is the first, will update these outputs,

improving on some of the gaps identified in the previous work and presenting new work based on outputs of a version of the SATIM energy systems model that is linked to an economic model (Merven et al., 2017). This first paper will focus on the multi-year calibration of the model using historical data and present a detailed picture of the national freight and passenger road transport system. The main improvement on the previous work is a longer historical window for calibration and more granular detail on heavy commercial vehicles so that the demand for freight (ton.km) could be calibrated to the national figure published by the Department of Logistics, University of Stellenbosch (Havenga, et al., 2016a). The data-rich picture presented is intended to support the development of projections of transport energy sector demand for infrastructure planning purposes, the compilation of greenhouse gas inventories, and the assessment of greenhouse gas mitigation measures, amongst other uses.

2. The demographics of the vehicle parc in South Africa

South Africa is made up of nine provinces of marked difference in size, population density and levels of economic activity, as shown in Table 1. Economic activity and car ownership are highly concentrated in small but densely populated Gauteng, for example, in contrast to the arid and sparsely populated Northern Cape. Three provinces – Western Cape, Gauteng, and Kwazulu-Natal with the country's biggest port, Durban – together account for 55% of the population, 69% of registered vehicles and 64% of gross domestic product. Much of this activity is concentrated in the cities of Cape Town, Johannesburg, Pretoria and Durban. This geography drives a large demand for transport, allowing the country to be described as having a 'spatially challenged economy' (CSIR, 2013).

The average total motorisation for South Africa, estimated for mid-2016 at 192 vehicles per thou-

Table 1: Demographics and motorisation of South Africa's provinces.

Province	Population (mid 2016) ¹	Total self-propelled vehicles (mid 2016) ²	Share of vehicles (%)	Motorisation (vehicle/1000 persons)	Contribution to national GDP (2015) ³ (%)	Land area share (%)
Eastern Cape	7 061 700	719 939	6.7	102	7.8	14.0
Free State	2 861 600	527 459	4.9	184	5.1	11.0
Gauteng	13 498 200	4 163 142	39.0	308	34.0	1.5
KwaZulu-Natal	11 079 700	1 477 700	14.0	133	16.0	7.7
Limpopo	5 803 900	604 318	5.6	104	7.2	10.0
Mpumalanga	4 328 300	733 541	6.8	169	7.5	6.0
Northern Cape	1 191 700	234 453	2.2	197	2.1	31.0
North West	3 790 600	528 044	4.9	139	6.5	8.7
Western Cape	6 293 200	1 721 222	16.0	274	14.0	11.0
Total	55 908 900	10 709 818	100.0	192	100.0	100.0

¹ = StatsSA (2016a), ² = eNaTiS (2016), ³ = StatsSA (2016)

sand inhabitants, now just exceeds the global average of 180 and is considerably more than the African average of about 44 (Organisation Internationale des Constructeurs d'Automobiles, 2016). But, even for the outlier of Gauteng, vehicle ownership is significantly lower than the 500 to 800 for developed countries and is more comparable to the BRICS counterparts of Russia and Brazil.

3. Modelling transport demand

In a bottom-up approach, energy consumption by any transport sector is directly driven by two factors: vehicle-km travelled, and conversion efficiency of the vehicle (referring to a road, rail or air vehicle). The vehicle-km travelled are in turn driven by the needs of society and the economy to move people and goods. Conversion efficiency depends mostly on the underlying technology, i.e., type of vehicle, fuel and vintage that make up the vehicle parc, and to some degree the patterns of utilisation of that technology. It is useful to treat passenger transport and freight transport separately, as the needs for moving people and goods have slightly different drivers and technologies.

(Armenia et al. 2010) proposed a detailed systems dynamics model, depicted by the causal loop diagram in Figure 1, to represent the demand for mobility and energy consumption of passenger transport. The model includes a number of drivers and interactions which define energy consumption in pas-

senger transport and illustrates the complex interactions and extensive data needs required to effectively model this sector. A diagram for road freight transport would be similar, in that fuel consumption is still the direct result of vehicle-km travelled and vehicle fuel efficiency. Several of the elements in Figure 1 are included in the calibrated vehicle parc model in the present study. These are: distance travelled per vehicle, total kilometres travelled, fuel consumption, fuel efficiency, total vehicle fleet, and average age of vehicles. Certain factors in Figure 1 affecting the vehicle-km travelled and fuel efficiency, such as traffic congestion, are difficult to quantify as they are not well understood locally. To compensate for this, the model was calibrated by adjusting the variables until the output matches the known fuel sales data. Once calibrated, the present study could be reasonably sure that the model returns realistic estimates of the number of operating vehicles and their annual distance travelled. By making an informed assumption regarding the average occupancies of different vehicle types, total private travel demand could be estimated.

4. Research methodology

SATIM is an energy-economic-environment systems modelling framework developed by the Energy Research Centre, University of Cape Town (Altieri, 2015; Energy Research Centre, 2015). It is developed according to ETSAP's TIMES modelling

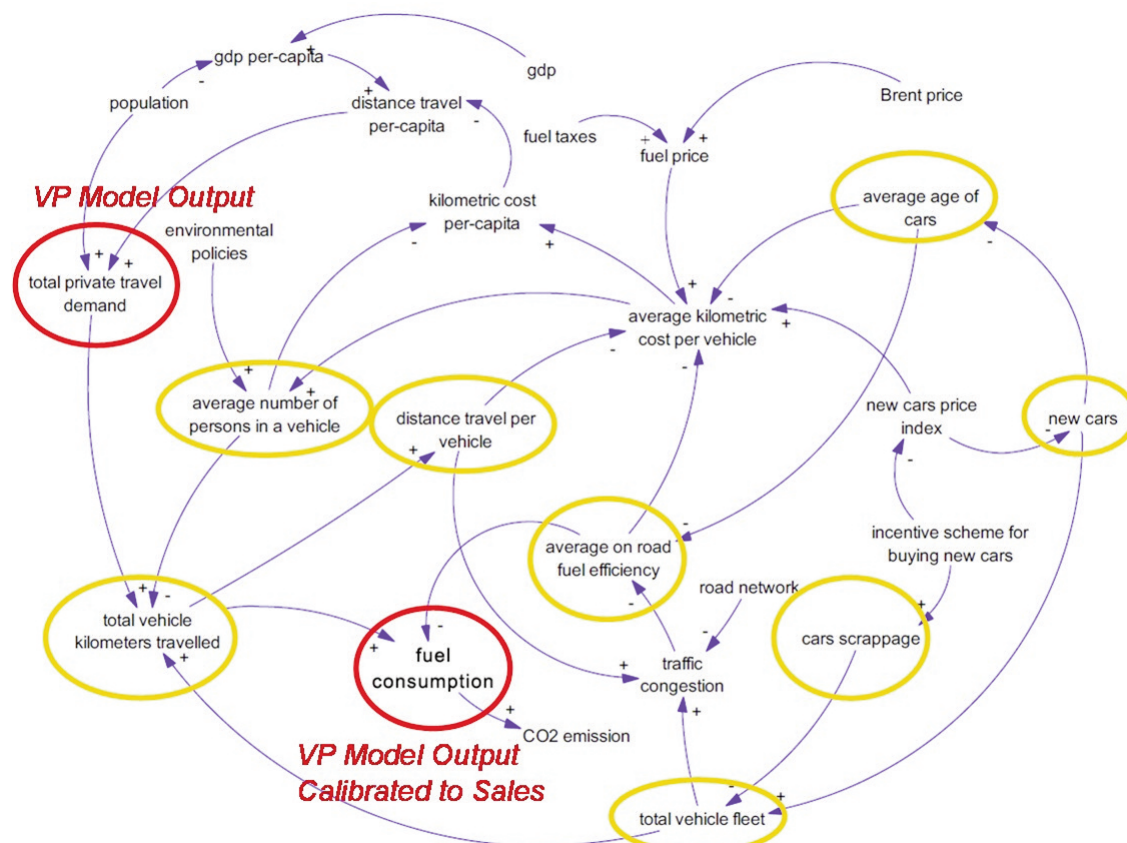


Figure 1: Causal diagram for energy needs for passenger transportation (Armenia et al., 2010).

framework, which has been linked to a computational general equilibrium model known as eSAGE. Both tools are developed using the general algebraic modelling system, but a number of supporting models developed on other platforms also support the modelling framework, as shown in Figure 2.

The present study focused on methodology, assumptions and outputs of the vehicle parc model component of the modelling framework shown in the highlighted textbox in Figure 2. The function of this model is to build a ‘bottom-up’ picture of the vehicle parc that can be used to generate a credible estimate of the following key variables to enable the energy system model (TIMES) to project energy demand forward:

- public and private passenger.km by mode/vehicle typology (e.g., gasoline minibus public or diesel car private);
- freight ton.km by mode/vehicle typology (e.g. gasoline light commercial vehicle (LCV)); and
- the stock, vintaged by age, of each typology and representative activity (annual mileage) and efficiency assumptions for each typology. These assumptions need to be calibrated so that the fuel demand of the model matches supply side data (fuel sales) as closely as possible.

The road vehicle parc is characterised by a long vintage window of around 30 years, given the high average age of stock in South Africa. The characteristics of new stock added and the activity levels of old stock can change annually. Thus, while calculations in this type of model are possible, the multiplication of many large arrays is required. Lumina’s Analytica platform (<http://www.lumina.com/>), an array-based modelling tool with a powerful visual interface, was selected for the first study (Merven et

al., 2012) for this reason. This vehicle parc model was updated for the present study and extended as follows:

- heavy commercial vehicles which were a single aggregate in the first study were disaggregated into nine vehicle typologies to assist with calibrating the model to the estimate of freight demand (ton.km) published by the Department of Logistics, University of Stellenbosch (Havenga, et al., 2016a);
- extension of the calibration window from seven to 14 years, spanning 2000–2013; and
- a parallel version in the open source R language was coded to aid collaboration; final calibration was performed in the R version.

4.1 Calculation and calibration

A schematic representation of the vehicle parc model and its data inputs and validations is shown in Figure 3. The procedure for calculation and calibration using the above parameters was broadly as follows:

1. Historic vehicle sales data collected by the National Association of Automobile Manufacturers of South Africa (Lightstone Auto, 2015) were adjusted by scrapping curves to develop an estimate of the stock of vehicles of different vehicle types for each model year, and the estimate was calibrated to the electronic national administration traffic information system (eNaTiS) registration database (eNaTiS, 2016) by adjusting the rate of scrapping.
2. Vehicle mileage estimates were developed for both passenger and freight vehicles, assuming that the annual mileage travelled by vehicles decays from an initial value as they age.
3. Fuel demand was calculated by multiplying the

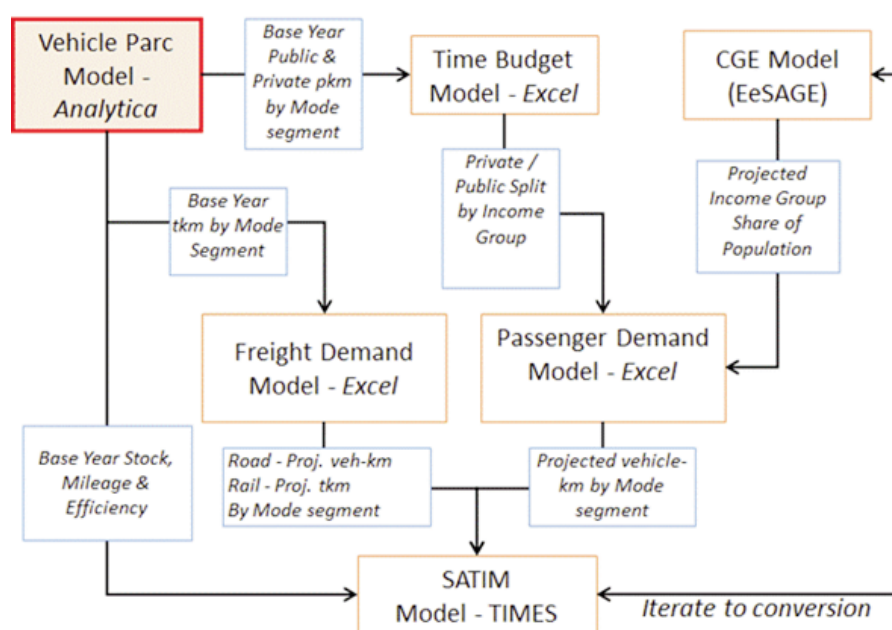


Figure 2: The SATIM transport modelling framework.

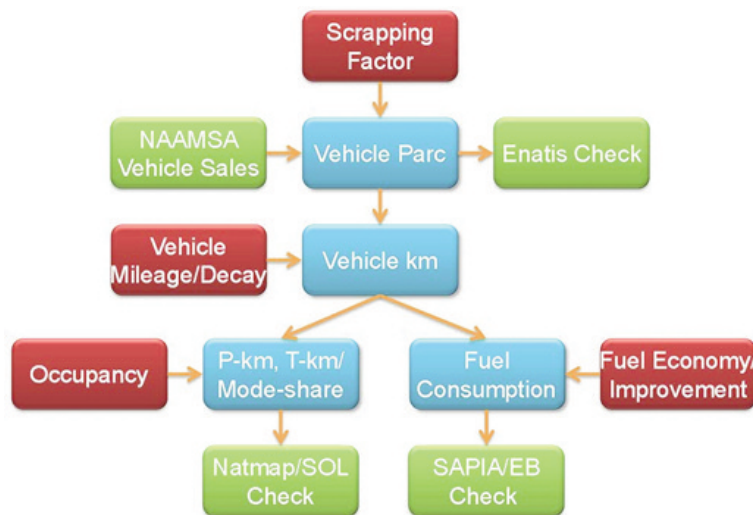


Figure 3: Schematic representation of the vehicle parc model and its data inputs and validations.

kilometres travelled, the fuel economy, and the number of vehicles for a vehicle typology. The typology fuel demands were summed to yield the vehicle parc fuel demand for gasoline and diesel and compared with aggregate fuel sales data.

4. An additional step for freight vehicles involved adjusting the average maximum capacity, capacity factor and mileage so that the model calibrated not only against fuel sales but also against published estimates of ton.km road freight demand found in the literature (Havenga, et al., 2016a).

4.2 Vehicle typologies and calibration targets

The typologies adopted for the vehicle parc model are presented in Tables 2 and 3, mapped to fuel type and more aggregate classifications. The aim for the model was to calibrate model fuel demand against data on road transport fuel sales across a window of 2000–2014. In the case of gasoline, 100% of sales were assumed to be used by road transport, but this was challenging in the case of diesel because, unlike gasoline, diesel is used for a wide range of off-road and stationary uses, including the fuelling of Eskom’s Ankerlig and Gourikwa power stations since 2007. The South African Petroleum Industry Association (SAPIA) and its members disaggregate fuel sales using a quasi-sector typology called trade categories, and this data is then collated and made available on request by the DoE (DoE, 2017). These categories offer some indication of what portion of diesel is used by road transport if an assumption of the share of road transport diesel can be made for each trade category. An assumed share of road transport in each trade category was adjusted iteratively until there was a relatively smooth trend in the shares of the sector demands over time, as shown in Figure 4. The resulting road transport share of diesel was

used to estimate a diesel calibration target for the model.

4.3 Other calibration aspects

4.3.1 Vintage profile

It is necessary to estimate the distribution of vehicles of different ages and technology levels in the parc, known as the vintage profile, to assess the impacts of new technologies entering the market such as on energy demand. The vintage profile can be determined by establishing a distribution of the probability of a vehicle surviving as a function of its age for each vehicle typology. Further detail on how this was done is provided in the supplementary file.¹

4.3.2 Vehicle mileage

The annual mileage of vehicles, when averaged over a large number, appeared to decay steadily from an initial value for each year of operation (Jackson, 2001; University of California at Riverside, 2002). This is important because it means that older, more-polluting vehicles would contribute proportionally less to transport demand than newer vehicles. This data is, however, not mandatory for capturing in the South African licence-renewal process and was not available. Mileage assumptions based on the United States Environmental Protection Agency’s Mobile6 model methodology (Jackson, 2001) were, therefore, adopted and scaled into the calibration process. Further detail on how this was done is provided in the supplementary file.

4.3.3 Fuel economy

The fuel economy of new vehicles was considered to decrease by 0.5% per annum between 2000 and 2014 for model implementation in the present study. This value was generally consistent with the more specific values in Europe during the period used in the present study. Details can be seen in the supplementary file.

Table 2: Typology of vehicle parc model mapped to aggregate categories: passenger vehicles.

Vehicle type (eNatis and the SATIM energy system model)	Fuel type	Vehicle parc model ID*
Passenger car	Diesel	CarDiesel
Passenger car	Gasoline	Gasoline
Passenger car	Diesel	CarHybridDiesel
Passenger car	Gasoline	CarHybridGasoline
Passenger car	Electricity	CarElectric
Bus	Diesel	BusDiesel
Minibus taxi	Diesel	MBTDiesel
Minibus taxi	Gasoline	MBTGasoline
Sport utility vehicle	Diesel	SUVDiesel
Sport utility vehicle	Gasoline	SUVGasoline
Sport utility vehicle	Gasoline	SUVHybridGasoline
Motorcycle	Gasoline	MotoGasoline

* These IDs are used in graphs and tables in the following sections.

Spread between 'light passenger vehicle' and 'light load vehicle' in eNaTiS data.

Table 3: Typology of vehicle parc model mapped aggregate: Freight vehicles.

Vehicle type	Weight typology (kg)	Fuel type	Model ID*
LCV	<3 000	Diesel	LCVDiesel
LCV	<3 000	Gasoline	LCVGasoline
MCV	3 000–7 500	Gasoline	HCV1Gasoline
MCV	3 000–7 500	Diesel	HCV1Diesel
HCV	7 501–12 000	Diesel	HCV2Diesel
HCV	12 001–16 000	Diesel	HCV3Diesel
HCV	16 001–20 000	Diesel	HCV4Diesel
HCV	20 001–24 000	Diesel	HCV5Diesel
HCV	24 001–32 000	Diesel	HCV6Diesel
EHCV	32 001–40 000	Diesel	HCV7Diesel
EHCV	40 001–50 000	Diesel	HCV8Diesel
EHCV	>50 000	Diesel	HCV9Diesel

LCV = light commercial vehicle; MCV = medium commercial vehicle; HCV = heavy commercial vehicle; EHCV = extra-heavy commercial vehicle

* These IDs are used in graph and tables in the following sections

Figure 4: Estimate of road transport share of diesel sales less Eskom consumption for calibration, where DOE EB: DoE's energy balances.

4.3.4 Occupancy and load factor

No published local empirical data was available to guide the deliberations for vehicle occupancy and load factor needed to calculate the demand for the model's passenger.km and ton.km. Initial freight load factors were drawn from the Road Freight Association's (RFA's) vehicle cost schedule (RFA, 2009) and then calibrated to the ton.km estimate for 2014 published by the Department of Logistics,

University of Stellenbosch (Havenga, et al., 2016a). The occupancy for passenger vehicle was taken from Merven et al. (2012). Details on load factors and occupancy can be seen in the supplementary file.¹

5. Results and discussion

The model achieved a generally good calibration for vehicle population and fuel demand and generated a number of statistics of interest for the South African vehicle parc. The aggregate calibration of the model with the registration database eNaTiS showed agreement within 3%, as shown in Figure

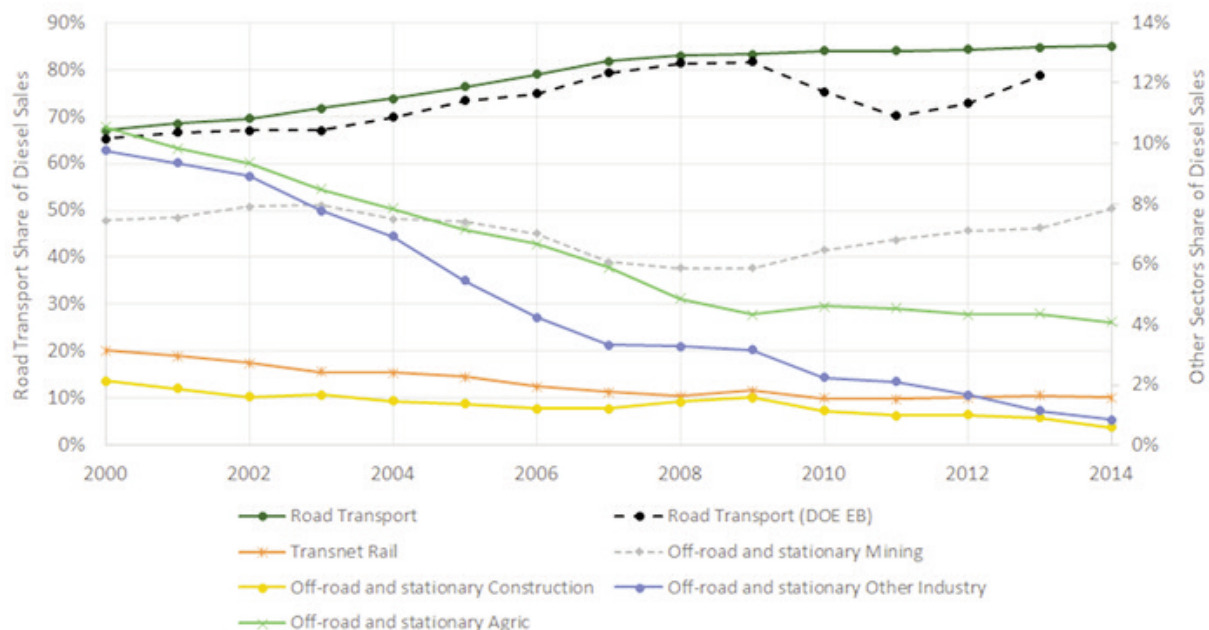


Figure 4: Estimate of road transport share of diesel sales less Eskom consumption for calibration, where DOE EB: DoE's energy balances.

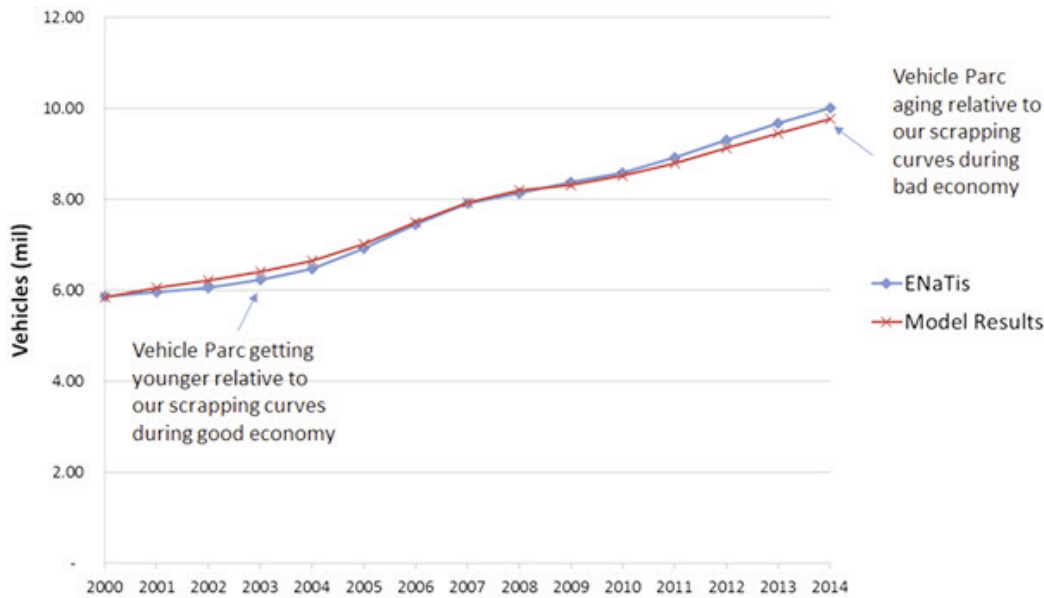


Figure 5: Calibrated vehicle parc model compared to the eNaTiS registration database.

5. This allowed the split of the vehicle parc into its typologies with some level of confidence, as presented in Tables 4 and 5.

The model indicated steady dieselisation of the light vehicle fleet between 2000 and 2014, as shown in Figure 6. The registration database (eNaTiS, 2016) has included a split between diesel and gasoline vehicles since 2015, and the model output compared well with this data.

The demand for passenger.km was calculated by multiplying the calculated vehicle.km with the assumed occupancy presented in Table 6 for each typology. The results show good agreement with other studies (NATMAP, 2005; NHTS, 2013). South Africa’s car-driving, high-income households and low-income public transport users live in peripheral

sprawl, so average trip distance is likely to be similar across modes, so motorised trip-based mode share and passenger.km mode share compare reasonably well.

Similarly, the demand for freight transport in ton.km was calculated by multiplying the calculated vehicle.km with the assumed load factors, as presented in Table 8, for each typology. In this case, however, the load factors were derived from a calibration process against a published figure of 231 billion ton.km for 2014 (Havenga, et al., 2016a). The detailed calibration results, including a split by vehicle typology and corridor, metropolitan and rural operating environments are presented in Appendix A in the supplementary file.

The energy intensity of road freight transport

Table 4: Vehicle typologies as a fraction of the passenger road vehicle parc for 2010.

Vehicle typology	Count of vehicles (2010)	Fraction (%)
CarDiesel	221 779	3.6
CarGasoline	4 832 681	78.3
CarHybridDiesel	0	0.0
CarHybridGasoline	1 113	0.0
CarElectric	0	0.0
BusDiesel	21 686	0.4
*MBTDiesel	23 911	0.4
*MBTGasoline	277 064	4.5
SUVDiesel	171 630	2.8
SUVGasoline	252 050	4.1
SUVHybridGasoline	541	0.0
MotoGasoline	372 534	6.0
Total	6 174 988*	100.0

* Total not calibrated to include the ‘Other self-propelled vehicles’ category in the eNaTiS registration database.

Table 5: Vehicle typologies as a fraction of the freight road vehicle parc for 2010.

Vehicle typology	Count of vehicles (2010)	Fraction (%)
LCVDiesel	852 698	33.2
LCVGasoline	1 394 106	54.2
HCV1Gasoline	6 061	0.2
HCV1Diesel	120 270	4.7
HCV2Diesel	31 074	1.2
HCV3Diesel	53 763	2.1
HCV4Diesel	12 249	0.5
HCV5Diesel	5 146	0.2
HCV6Diesel	74 175	2.9
HCV7Diesel	19 406	0.8
HCV8Diesel	434	0,0
HCV9Diesel	514	0.0
Total	2 569 897*	100.0

* Total not calibrated to include the ‘Other self-propelled vehicles’ category in the eNaTiS registration database.

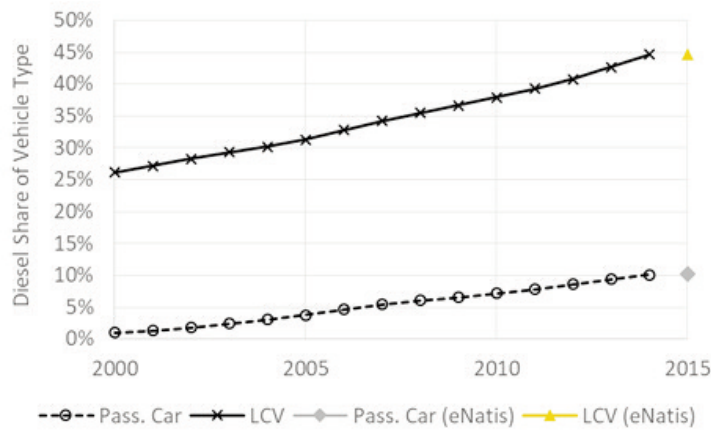


Figure 6: Modelling trend in dieselisation of the light vehicle fleet compared with the registration database for 2015.

Table 6: Model-generated passenger transport data for South Africa (2014).

	Total vehicles (1000 vehicles)	Total vehicle-km (billion vehicle-km)	Km per vehicle (1000 km)	Occupancy (person/vehicle, %)	Activity (billion p.km)	Modal share (2014) (%) of p.km	NATMAPa (2005) (%) of motorised trips	NHTSb (2013) (%) of motorised trips
<i>Public</i>								
Large bus	23	0.61	26.1	25	15.2	6	10	8
MBT	278	7.40	26.6	14	103.7	38	40	35
Train ^c					13.9	5	8	5
Sub-total					132.7	49	58	48
<i>Private</i>								
Pass. car	5722	83.5	14.6	1.4	116.9	43.0		52.0
SUV	644	12.8	19.9	1.4	18.0	7.0		
M/cycle	478	3.6	7.6	1.1	4.0	1.5		0.4
Sub-total					138.9	51.5	40.0	52.4
Grand total					271.6			

(a) Source: DoT, 2009

(b) Source: Stats SA, 2013 – Calculated by mode taken on allocated travel day - NOT stated mode preference.

(c) Train data from literature, not the model - Intra city data only for 2006/2007 (Metrorail, 2007). Data for inter-city is not published by the respective vendors

Pkm = passenger, NATMAP = National transport masterplan, NHTS = National household travel survey, MBT = Minibus taxi.

was estimated to be 1.9–2.0 MJ/ton.km by dividing the estimated diesel sales to road transport (the calibration target) and the model's forecast of road freight diesel demand by the demand in ton.km. The result compares favourably with international data for similar markets, such as Australia (shown as AUS in Figure 7).

Figure 8 shows that the model had excellent agreement with both gasoline and diesel use during the calibration years 2000 to 2014, with the exception of gasoline in 2014, thus providing validation for the model and the indicators generated from it.

In the version of the model developed for the present study, the calibration period was extended compared to the period used in Merven et al. (2012), resulting in a relatively less-close fit to real world data, but still within acceptable bounds for this type of model. The standard deviation of errors was 4.8% for both gasoline and diesel and was well

under 10%, except for the outliers in 2014 (11%). The parameters input to the model in this study as compared with Merven et al. (2012) can be contrasted using Phase 1 column in Tables and in the tables in the supplementary section. A comparison for gasoline cars only is presented in Table 7.

Both fuel economy and mileage in the present study are about 10% lower, which may reflect an improved estimate because of the longer calibration window, but to some extent the differences also arise from the pressure on the calibration of outliers in demand in 2008, 2013 and 2014. The possible causes for this are discussed in more detail below. The data presented in Table 13 of the supplementary file implies that the fuel consumption improvement was relatively low at 0.5% per annum over the study period. The population of gasoline vehicles, however, still grows at over 3% per annum, despite dieselisation. If vehicle mileage was con-

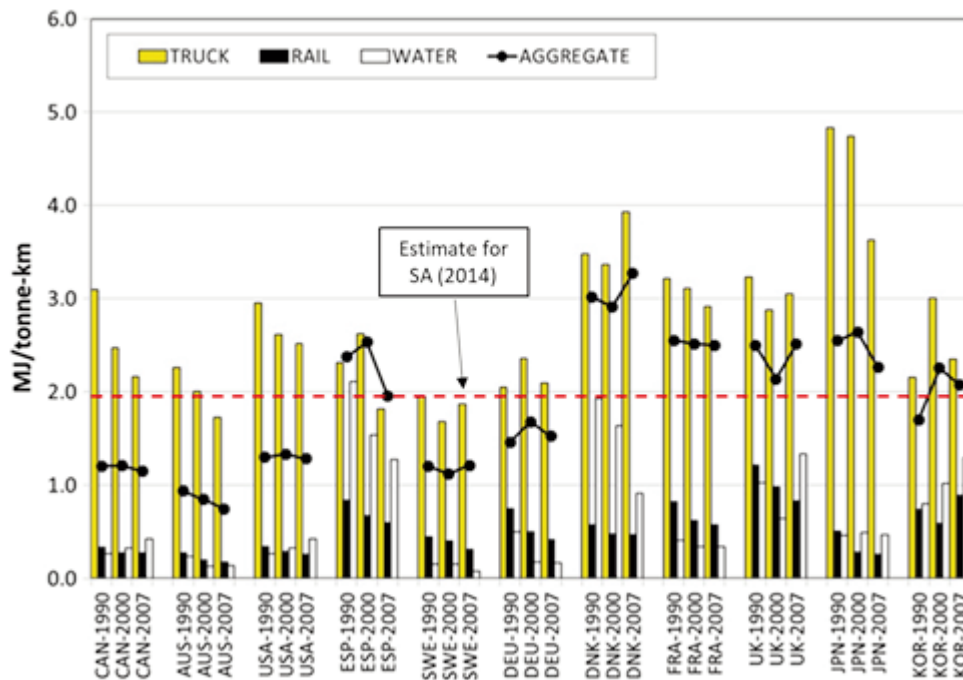


Figure 7: South African road freight (truck) energy intensity in 2014 compared to selected OECD markets 1990–2007 (Eoma et al., 2012, Havenga et al., 2016a)

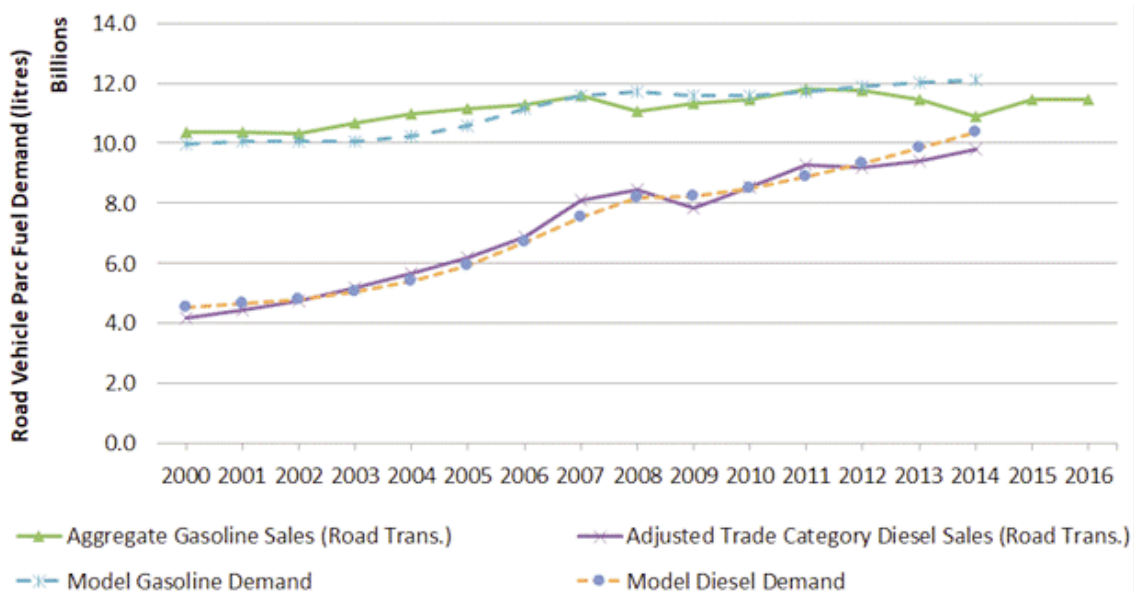


Figure 8. Model’s fuel demand vs actual fuel consumption for 2000 to 2014.

Table 7: Comparison of calibration parameters for gasoline cars between Merven et al. (2012) and the present study.

	Year	Fuel economy (l/100 km)		Annual mileage (km)	
		New	Fleet average	New	Fleet average
Phase 1	2006	8.3	9.1	24 000	16 169
Present study	2014	7.8	8.2	21 000	14 457

stant, as assumed by the model, future growth in gasoline demand would be sustained, as shown in Figure 8. It can, therefore, be deduced that the recent drop in demand observed in the gasoline sales data is driven by consumers travelling less, while recognising the difficulty of verifying this without empirical investigation. Most notably, local

authorities do not enforce its capture and the eNatis does not make the limited dataset that has been accumulated from partial completions available, despite the annual vehicle registration form historically having a placeholder for vehicle mileage. The assumption of 0.5% fuel consumption improvement qualifies that the sources of error in the

model's fuel demand relative to observed sales is likely to include the following:

- short- and long-run elasticity of demand in response to fuel price variations and average income due to consumers travelling less;
- other effects on demand for private transport like changing urban form and fluctuating congestion; and
- the error in the fuel sales data because of changes in parties responsible for data collection, compounded by changes in the structure of the gasoline-fuels value chain.

The model is intended for passing parameters to a long-term projection model. It does not attempt to integrate price or income elasticities and assumes that consumer preference for annual mileage is constant over the period for a given vehicle age. Considering that the vehicle population grew monotonically over the study window, any significant drop in demand in response to price increase would, therefore, result in significant error. The model responds indirectly to low economic growth through reduced car sales, but, as seen in Figure 5, this effect is relatively dampened compared to the fuel demand volatility seen in Figure 8. The periods of error in the calibration correspond to price volatility and economic growth variations as shown in Figure 9, with gasoline seemingly more responsive to price fluctuations, and diesel more responsive to GDP/capita variations.

The levelling off of GDP/capita caused by a struggling economy also seems to have depressed demand for gasoline in a longer-run effect relative to the calibration from 2012 onwards, presumably resulting from lower vehicle activity. Figure 8 shows

that gasoline demand recovered in 2015 and 2016 in response to significant real price drops, while diesel demand remained low.

Travel time data suggests that congestion and trip distances have increased in South Africa (SEA, 2017), so another possible reason for the observed drop in demand is data error. Historically, liquid fuels were supplied by a regulated oligopoly of the oil majors who collected detailed demand data to enable the complex distribution of fuels. The industry has changed as a result of competition legislation and legislation to enable access to the value chain by historically disadvantaged entrants. This resulted in the responsibility for data collection shifting from one of the majors – firstly to the industry association, SAPIA, and then to the DoE. At the same time a number of independent wholesalers emerged, including Afric Oil, Gulfstream Energy, Mzumbe Oil, Women of Africa Fuel and Oils, Siyanda Petroleum, and Yem Yem Petroleum. These now deal in large volumes in some cases, building on a base of procurement by the state and state-owned enterprises (Transport World Africa, 2014; Greve, 2013). This is to the extent that, while independent wholesalers traditionally were not directly surveyed for the national statistics, they are in some cases believed to be importing fuel independently in large enough volumes to introduce significant statistical uncertainty into the national energy balance (DoE, 2017b). The DoE has, therefore, not only taken on a complex statistical function, but performs this function at the time when the industry is rapidly becoming more complex and difficult to survey. It seems likely, then, that the levelling-off of gasoline demand is temporary and that economic recovery and ongoing improvement in data collec-

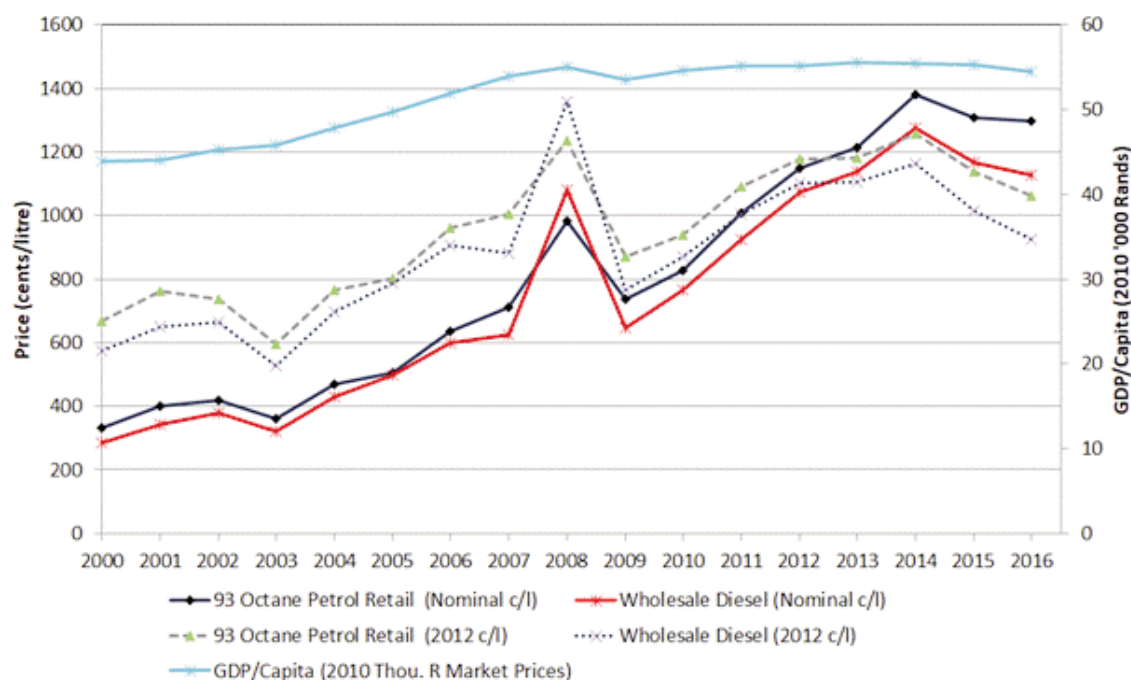


Figure 9. Gauteng gasoline and diesel prices and GDP/capita, 2000–2014.

tion methods may see the official figure for gasoline demand rise again by the order of 2–3% per annum until there is a significant penetration of real-world, low-fuel economy vehicles into the vehicle parc.

A key aim of this study was to improve on previous work by calibrating freight demand from the model with that published by the Department of Logistics, University of Stellenbosch. The number of freight vehicle typologies was increased and load factors adjusted, taking some account of the limited consultation (Havenga & Simpson, 2016b) that was possible with the Department of Logistics within the limitations of the present study. A data-rich output was a calibrated split of ton.km by vehicle typology and by region (corridor, rural and metropolitan), presented in Table 8.

The verified ton.km figure further enabled an estimate of freight energy intensity for the country, as presented in Figure 7. This calculation, however, required an estimate of the diesel sales to road transport as a calibration target and this can be considered uncertain given that it was derived from a trend-smoothing exercise using sales data disaggregated by ‘trade category’, a legacy classification that gives limited guidance on the commercial activity the fuel was used for. The DoE is in the process of moving to International Standard Classifications in its questionnaires (DoE, 2017b), which may enable more certain estimates in the future. A further uncertainty was that the diesel demand by Eskom was notably high during 2010–2015, and this must be accounted for in the trend analysis. In addition, the energy balances for those years (as published at

the time of writing) suggest that the Eskom consumption was excluded from trade category data, the total of which is equal to total final consumption excluding transformation. Liaison with the DoE (2018), however, confirmed that the Eskom consumption is indeed included in the trade category data and this was, therefore, adjusted downwards by the additional amount before the share of road transport was estimated. Improved statistical methods will, however, only partly reduce uncertainty. There is no substitute for empirical sector studies and far more needs to be understood about energy use in the agriculture, construction and mining sectors, if there should be any certainty that the residual diesel in the calibration reflects that used by stationary and off-road activities.

The new freight data presented by the present study, however, represents a rare attempt to achieve agreement in key parameters across modelling efforts by different teams in related fields, and will hopefully be an example for improved collaboration, more effective validation, and better support for policy and planning in the energy and transport spheres.

6. Conclusions

A vehicle parc model for South Africa incorporating detailed estimates for efficiencies and activity levels for a variety of vehicle typologies was developed and calibrated against national sales of gasoline and diesel over 15 years, from 2000–2014. The model has 11 freight vehicle typologies that were used to develop a parallel calibration of the ton.km demand

Table 8: Freight demand calibration output by vehicle typology and operating environment.


SATIM vehicle type	Load factor (ton/veh)	Assumptions (%)			Model freight demand (billion ton.km)				
		Share of ton.km that is corridor	Share of ton.km that is metro	Share of ton.km that is rural	Total	% share of demand	Corridor	Metro-politan	Rural
HCV1Diesel	1.4	0.4	86	14	7.77	3.4	0.03	6.67	1.07
HCV1Gasoline	1.5	8	75	17	0.06	0.0	0.00	0.04	0.01
HCV2Diesel	2.5	10	54	36	2.38	1.0	0.24	1.28	0.86
HCV3Diesel	5.0	24	62	14	16.19	7.0	3.81	10.11	2.27
HCV4Diesel	7.9	38	40	22	6.47	2.8	2.46	2.61	1.40
HCV5Diesel	10.9	53	31	16	3.96	1.7	2.11	1.22	0.63
HCV6Diesel	20.7	54	24	22	142.10	61.4	76.52	34.18	31.41
HCV7Diesel	18.6	55	16	29	31.51	13.6	17.29	4.97	9.26
HCV8Diesel	18.4	57	15	28	1.24	0.5	0.71	0.18	0.35
HCV9Diesel	21.9	57	17	26	0.70	0.3	0.40	0.12	0.18
LCVDiesel	0.4	0	84	16	8.89	3.8	0.03	7.46	1.40
LCVGasoline	0.4	0	82	18	9.98	4.3	0.02	8.19	1.78
<i>tkm calibration comparison</i>									
Total					231.3	100.0	103.6	77.0	50.6
Dept of Logistics (Havenga, et al., 2016a)					231.0		103.4	77.0	50.6
Calibration error (%)					-0.11	0.20	0.00	0.12	

of the model with that estimated by the freight demand model of the Department of Logistics, University of Stellenbosch for 2014 (Havenga, et al., 2016a). This enabled an estimate of the energy intensity of freight transport for the country (1.9–2.0 MJ/ton.km) to be made, one that can be used to benchmark the energy efficiency of the freight logistics industry. The model output furthermore provides a data-rich picture of the activity levels, efficiencies and contribution to meeting passenger and freight demand of different vehicle typologies.

Some interesting trends emerged from the time series of input and output data, as follows:

- Steady dieselisation of the light vehicle fleet has been occurring.
- The fuel economy of the light vehicle fleet has been improving only very slowly, if at all.
- The consumption of gasoline in particular has dropped off steadily since 2011 and seems to relate to lower activity levels, driven by economic factors.

The following important data issues emerged from the study:

- In general, the quality of energy related data received was of concern in all the major sources: fuel sales and registration statistics collated by government in partnership with industry, and vehicle sales data collated by industry and sold by a private concern as proprietary data.  following were key issues:
- Metadata is sparse or non-existent.
- Obvious validation checks have sometimes not been performed – for example, the sum of disaggregates of the same commodity might not match or time series of quantities have implausible step changes or trends.
- A poor understanding of the technical details is sometimes apparent – for example, the difference between CO₂ emissions and emissions standards for local air pollutants.
- Older data has been removed from the public record, so that developing time series is difficult.
- The sparseness of activity data – for example, annual mileage over the life of the vehicle or vehicle occupancy and load factors necessitated many assumptions.
- Diesel use in the agriculture, construction and mining sectors needs to be better understood in order to make allocations of diesel use to sectors for modelling and greenhouse gas inventory purposes with any certainty.

This study suggested that better energy policy and planning going forward requires stakeholders to collaborate to improve the quantity, quality and accessibility of energy and environment data.

Transport planning and planning for energy for transport are particularly high on the national agen-

da, with congestion in cities increasing, public transport networks expanding at great expense, and the costs of energy imports rising. The data-rich picture provided by this model is, therefore, a useful input to many policy activities other than the projection of energy demand. For example, extensions of this type of model are particularly useful for rapid assessment of the impacts on demand of disruptive transport technologies, including battery electric vehicles, hybrid electric vehicles and hydrogen fuel cell vehicles. This has relevance to the large gasoline fuels sector in the country, which could be severely affected by penetration of these technologies. Future work will aim to explore these impacts.

Note

1. Supplementary material can be found at [TO BE INSERTED BY LAYOUT]

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