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Approaches and techniques for modelling CO₂ emissions from road transport

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Abstract

Transport accounts for around a quarter of CO₂ emissions globally. Transport modelling provides a useful means to explore the dynamics, scale and magnitude of transport related emissions. This paper explores the modelling tools available for analysing the emissions of CO₂ from transport. Covering a range of techniques from transport microsimulation to global techno-economic models, this review provides insight into the various advantages and shortcomings of these tools. The paper also examines the value of having a broad range of perspectives for analysing emissions from transport. The paper concludes by suggesting that the broad range of models creates a rich environment for exploring a spectrum of policy questions around the emissions from transport, and the potential for combining modelling approaches further enhances the understanding that can be attained.

Key words: CO₂ Emissions; Transport Models; Sustainability; Road Transport

Introduction

The emission of greenhouse gases (GHG) and the consequential climate change impacts is one of the greatest challenges facing the global community [Stern, 2007]. Of the GHGs causing climate change, CO₂ is the most significant and the transport sector accounts for around 25% of CO₂ emissions, although this varies by scale and location, with energy supply, business and the residential sector making up the rest [DECC, 2013, International Energy Agency, 2009]. In order to combat rising CO₂ emissions it is essential to have a clear understanding of the generation of emissions within the transport system. The transport system is defined in this work as the
infrastructure facilitating the movement of people and goods. This can include the
physical infrastructure, ICT and vehicles. The policy and regulatory environment is
also included in the review of modelling approaches as an important aspect of the
transport system. This paper focuses on emissions from road transport,
predominantly passenger transport rather than freight specifically, although freight
may be implicitly included within some of the modelling techniques reviewed.

Modelling transport is a key element of transport research, exploring a multitude of
factors and situations at a range of scales to provide insight and understanding of
the challenges that face the infrastructure and the stakeholders who manage the
infrastructure [Hensher and Button, 2008]. Motivations behind modelling are as
numerous as the challenges, and include such factors as improving inefficiencies in
networks and examining the potential traffic impacts of changes to infrastructure
[Gudmundsson, 2011, Brömmelstroet and Bertolini, 2011], exploring socio-economic
effects of transport [Wismans et al., 2011], understanding supply and demand for
transport [Proost and Van Dender, 2011], as well as modelling air quality and
emissions [Brand et al., 2012, Samaras et al., 2012]. The scale of transport
modelling is also wide-ranging, from the micro level modelling of a single
intersection, or local road network, to the global transport system.

Transport modelling provides tools that inform the user about the system in question
and should aid in planning and policy decisions [de Dios Ortúzar and Willumsen,
2011]. In order to facilitate a transition to a sustainable transport system, the choices
that are made now must take into account the necessary emission reductions to
prevent dangerous climate change and meet emission reduction targets
[Intergovernmental Panel on Climate Change, 2007]. There are a range of
challenges arising from the transport sector, not least environmental damage, and
examination of these issues through the use of transport models can aid decision
makers in dealing with these challenges [de Dios Ortúzar and Willumsen, 2011].

There are many examples of how and where emissions from transport have been
accounted for through modelling approaches. At the local scale, traffic network
models can provide planners with information about the emissions impacts of
infrastructure decisions by predicting traffic flows and travel decision making
[Gudmundsson, 2011]. There are also behavioural models [Stern and Richardson,
2005] that can provide additional insight on travel behaviour [Hensher et al., 2013]
and the potential emissions impacts of changes to infrastructure, such as the uptake
of electric vehicles [Shafiei et al., 2012b]. At a global level, large scale energy
systems models [e.g. Kim et al., 2012] can provide information about the transport
sector internationally. These can support understanding of the action that is required
at transnational level to mitigation emissions of CO$_2$ in order to prevent dangerous climate change and thereby inform policy at the international and national level.

This paper will provide a review of the modelling approaches available for accounting for emissions of CO$_2$ from road transport, across the different scales. While previous work has reviewed transport modelling approaches, the quantification of emissions from transport in models has not been explicitly emphasised and this paper aims to address this gap. The paper will begin by providing an overview of the modelling approaches that are considered in this work and the reasons for their inclusion. The models will then be analysed and critiqued, with example packages and capabilities presented. A synthesis will be presented that draws together the understanding of the different approaches and techniques that has been presented in the preceding sections and offers proposals on the relative merits of these models. This is followed by some final conclusions.

**Model classification for review**

The following types of models are reviewed in this research:

- Traffic network models: microsimulation;
- Behavioural models;
- Agent based modelling;
- System dynamics modelling;
- Techno-economic, and;
- Integrated assessment models

This classification of models has been used to provide a broad perspective of the ways that the transport system can be modelled. The models have been classified as such because of the different techniques that are used within the approach or the spatial and temporal context of the model. There are a spectrum of geographical and temporal scales represented, from the local to the global and the near term future to a forecasting horizon of 2100. Figure 1 shows the scales covered by the modelling approaches included in this paper. The models may have differing underlying epistemologies yet they all have an ability to represent elements of the transport system. They are also able to examine similar policy questions, particularly the question of the scale of emissions from transport, despite their differing perspectives and geographical scope. The various modelling tools they contain may be used in different ways by a variety of practitioners in order to explore a range of transport related challenges. It is therefore interesting to consider the relative advantages and shortcomings of the different types of model and the necessity for multiple approaches in modelling the transport sector.
This review will examine these issues by seeking to answer the following questions:

1. Which techniques and tools are used to model transport systems?
2. How do these capture emissions from transport? and,
3. What are the advantages and disadvantages of the different approaches?

Figure 1 - Spatial and temporal scales for the models reviewed
# Systematic Review of Transport Modelling Approaches

## Table 1 - Review of approaches for modelling emissions from road transport

<table>
<thead>
<tr>
<th>Modelling Approach</th>
<th>Example packages/studies</th>
<th>Elements of transport system captured</th>
<th>Underlying concepts</th>
<th>Emissions calculation</th>
</tr>
</thead>
<tbody>
<tr>
<td>1. Traffic Network models - Microsimulation</td>
<td>• DRACULA (Dynamic Route Assignment Combining User Learning and microsimulation) • VISSIM</td>
<td>Captures movement of vehicles within a pre-defined traffic network. Aggregate data across the network for emissions, delays, travel time etc.</td>
<td>Microsimulation modelling is built on the principles of the four stage trip model (de Dios Ortúzar and Willumsen, 2011). The four stage model is as follows: 1. Trip generation 2. Trip distribution 3. Mode split 4. Traffic assignment</td>
<td>DRACULA assumes that fuel consumption factors are constant for vehicles that are idling or decelerating and calculates fuel consumption for accelerating vehicles using equation 1. This is then converted into CO₂ under the assumption that fuel burned is converted to CO₂. This does not take into account alternatively fuelled vehicles. VISSIM: requires an add-on to module ‘enViVer Pro’ which imports vehicle data to calculate CO₂ emissions in the study area and outputs a table and graph of this.</td>
</tr>
<tr>
<td>2. Behavioural models</td>
<td>• Stern and Richardson (2005) process-oriented</td>
<td>Captures greater detail about individual decision making a travel choices</td>
<td>Activity based models explore the activities that necessitate mobility to understand levels of transport and</td>
<td>Quantifying levels of transport demand that can then result in calculation of emissions of CO₂.</td>
</tr>
</tbody>
</table>
framework, see Figure 2
generate a clearer picture about
decision making. Travel is seen a
result of activities and behaviour.

<table>
<thead>
<tr>
<th>3. Agent Based modelling</th>
</tr>
</thead>
<tbody>
<tr>
<td>• MATSIM (Multi-Agent Transportation Simulator Toolbox)</td>
</tr>
<tr>
<td>Captures the dynamics of behaviour in the transport system, as for behavioural models above, captures more detail than just the trips that are made but understands behaviour and motivation</td>
</tr>
<tr>
<td>Models behaviour through a series of heterogeneous agents</td>
</tr>
<tr>
<td>Coupled with an emissions model can deliver unique insight into travel behaviour and road transport emissions</td>
</tr>
<tr>
<td>MATSIM itself does not calculate emissions from road transport. However, [Hatzopoulou et al., 2011] combined outputs from MATSIM with MOBILE6.2 to calculate emissions</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>4. System Dynamics</th>
</tr>
</thead>
<tbody>
<tr>
<td>• ASTRA – Assessment of Road Transport Strategies [Rothengatter et al., 2000]</td>
</tr>
<tr>
<td>• MARS – Metropolitan Activity Relocation Simulator [Research Centre for Transport Planning and Traffic]</td>
</tr>
<tr>
<td>Can capture interactions and feedbacks within complex systems such as transport</td>
</tr>
<tr>
<td>System dynamics models are built on causal loop diagrams and stock and flow relationships</td>
</tr>
<tr>
<td>ASTRA: Environment Sub-module calculates gaseous transport emissions. Figure 3 shows the structure of this calculation within the sub module, it includes factors for cold start, hot emissions factors etc.</td>
</tr>
<tr>
<td>MARS: 3 separate equations calculate emissions from road transport, the first calculates emissions from specific cars using speed output from MARS, another, total CO$_2$ for private cars using distance and number of trips from MARS outputs</td>
</tr>
</tbody>
</table>
### 5. Techno-economic models

<table>
<thead>
<tr>
<th>Models</th>
<th>Methodology</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Roadmap</td>
<td>Capture large scale dynamics of the transport system through the socio-economic changes that occur, using a top-down modelling approach</td>
<td>Draws on socio-economic characteristics and forecast changes to estimate transport demand and technological forecasts are used to develop emission factors that allow projections of future emissions to be established. Roadmap: runs over 5 year time steps from 2000 to 2015.</td>
</tr>
<tr>
<td>World Energy Projection System Plus (WEPS+)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>(EIA), 2011</td>
<td></td>
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</table>

**Roadmap:** Tank-to-wheel CO\(_2\) emissions are estimated based on energy consumption, calculated from vehicle km travelled (VKT), which is established using socio-economic indicators, and vehicle efficiency using the following equation:

\[
\text{Energy consumption (MJ)} = \text{vehicle efficiency (MJ/KM)} \times \text{VKT}
\]

Figure 4 also shows how the emissions calculations are structured in Roadmap.

### 6. Integrated Assessment models

<table>
<thead>
<tr>
<th>Models</th>
<th>Methodology</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Global Change Assessment Model (GCAM)</td>
<td>Large scale modelling of economy and environment with a sub-module for transport, capturing technological change and environmental impact driven by socio-economic factors</td>
<td>Runs in time steps of 15 years from 2005 to 2095 and achieves regional equilibrium in the regional markets in each time step.</td>
</tr>
<tr>
<td>(Kim et al., 2012)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Demand for transport is calculated in GCAM using equation 1. Equation 2 calculates the costs of passenger transport modes and Equation 3 uses this to calculate mode share. These can then be coupled with emission factors to provide regional and global projections of emissions from transport.
Synthesis of Modelling Capabilities by Model Type

The models that have been included in the systematic review in Table 1 represent a spectrum of modelling approaches spanning the spatial and temporal scales that are shown in Figure 1. The example packages were selected as they are representative of the techniques used in modelling the transport system and demonstrate different approaches to quantifying emissions from road transport. The following section synthesises the findings from the table and draws insights to the advantages and disadvantages of the various approaches. The traffic network, behavioural and agent based modelling approaches have been grouped together, as these represent the micro-scale approaches which deal with disaggregate and capture individuals' or vehicles' movements, choices and behaviours. These are followed by the system dynamic modelling approaches and the macro scale techno-economic and integrated assessment models. Examples of combining modelling approaches are also provided.

Traffic Network, Behavioural and Agent Based Models

Traffic network models cover a range of scales, from macro to micro. This review has focused on microsimulation as it represents a key subset of these models. Transport demand is generated within microsimulation modelling as follows. An origin-destination (O-D) matrix is used with the models to represent individual trips within the network and in the traffic assignment part of the four stage model [McNally, 2008]. The network that underlies traffic models is constructed from real world traffic networks using a series of 'nodes', 'links' and 'zones' [Willumsen, 2008]. Nodes represent junctions and intersections whilst links are roads that connect nodes. The zones are areas between which (or within which) trips are made and from which the O-D matrix is constructed. For example, a trip will be from origin (O) – zone x, to destination (D) – zone y [Atkins-ITS, 2013]. Further details on the behavioural rules that govern microsimulation of traffic can be found in the detailed manuals for the models [Liu, 2007; Verkehr, 2011]. Table 1 summarises the general method for emissions calculation within microsimulation packages, whilst Equation 1 below shows the calculation of fuel consumption in DRACULA, which is then used to calculate CO$_2$ emissions [Liu, 2005].

Calculation of fuel consumption in DRACULA:

\[ f = c_0 + c_1 \cdot a \cdot v \]  

Equation 1

Where:

\( f \) = fuel consumption factor

\( v \) = speed of the vehicle
a = acceleration of the vehicle

and \( c_0 \) and \( c_1 \) are constants defined in Ferreira, 1982.

For VISSIM, an additional add-on package, ‘enViVer’ is required for the calculation of emissions PTV Group, 2015. This package draws on the emission calculations developed in the VERSIT+ model which uses the following equation to calculate traffic emissions for pollutant \( j \) (\( TE_j \), g/hour):

\[
TE_j = \sum_{k,m} \left( E^F_{j,k,l} \times TV_{k,m} \times L_m \right)
\]

Equation 2

"Where \( E^F_{j,k,l} \) presents the predicted mean emission factor (g/km) for pollutant \( j \), vehicle class \( k \) and speed-time profile \( l \), \( TV_{k,m} \) presents the traffic volume with respect to vehicle class \( k \) (vehicles/hour) for a particular section of road \( m \), where speed-time profile \( l \) would apply and \( L_m \) presents the length of road section \( m \) (km)" [Smit et al., 2006, p17]. This is used to calculate the emissions across a network simulation in VISSIM, accounting for different vehicle classes and the speed-time profile for each vehicle.

By understanding the levels of demand for mobility within the traffic system and according to the scenario under scrutiny, the level of emissions can be established and the impacts of policy changes analysed Dowling et al., 2002. There are a range of additional behavioural rules within microsimulation that are not identified in column 4 of Table 1 ('Underlying Concepts') and details of these can be found in the individual model manuals Liu, 2007, Verkehr, 2011. Overall, microsimulation provides a series of useful techniques for modelling emissions from road transport, as it allows small changes in the network to be analysed in terms of their impacts on transport and the resulting emissions.

One of the goals of behavioural research in transport is to understand how travellers use the transport infrastructure in order to better predict needs and enhance decision making, thus providing a more holistic view of activities Davidson et al., 2007, Stern and Richardson, 2005. Behavioural models draw heavily on the disciplines of social psychology and behavioural economics for the underlying principles and frameworks Schaap and van de Riet, 2012. Stern and Richardson 2005 present a ‘process-oriented framework’ for understanding travel behaviour, as shown in Figure 2. This shows how additional factors beyond the need to travel from and origin to a destination can be considered in understanding travel behaviour, including factors such as the motivation for travel, scheduling and any constraints on travel.
Activity based models focus on the underlying motivations behind travel decisions, allowing travel patterns to be more clearly understood and to quantify the demands on the mobility system [Axhausen and Gärling, 1992, McNally and Rindt, 2008]. By quantifying the amount of travel within a network through an activity based approach, emissions factors can be applied to estimate the emissions of CO$_2$ from road transport.

Figure 2 - Process-oriented framework of the behaviour of road users (adapted from Stern and Richardson, 2005)

Agent based models have been used in activity based approaches, but there are also elements of agent based modelling principles incorporated within some microsimulation models. Agent-based modelling characterises a series of agents within a practice space, the modelled world in which the agents interact, in order to understand behavioural dynamics [Köhler et al., 2009]. Agents have behaviours and
interact with other agents and their environment, which can alter their behaviour and this provides insight into the dynamics of the system being modelled [Macal and North, 2010]. These approaches have been used in the social and biological sciences to understand behaviour [Grimm et al., 2006] and there are examples of their use in transport, including including Köhler et al., [2009] Quadstone Paramics [2013] and Shafiei et al (2012a). There are also examples of the inclusion of agent-based modelling principles in traffic micro-simulation models. PARAMICS is a traffic network model that characterises pedestrian activities and pedestrian traffic interactions, thereby incorporating an agent based modelling approach [Quadstone Paramics, 2013]. In addition, qualitative methods and virtual reality approaches outlined by [Dougherty et al., 2000] are areas that have the potential to increase the degree of agency within traffic network simulation. An additional advantage of using agent based modelling to explore systems such as those found in the transport sector is that a series of ‘heterogeneous agents’ can be modelled. These allow detailed insight into the interactions in the system and a ‘bottom-up’ approach to be taken [Garcia, 2005, Shafiei et al., 2012a].

MATSIM captures the dynamics of the transport system with a behavioural oriented approach, characterising this through agents within a traffic network. The MATSIM programme itself does not calculate road transport emissions, but a study by Hatzopoulou et al., [2011] coupled MATSIM with the MOBILE6.2 model to estimate the emissions generated by the travel behaviour being modelled. MOBILE6.2 uses fuel economy outputs to calculate emissions of CO\(_2\) but these are not varied for speed and engine type, as the calculations for other pollutants in MOBILE6.2 are, however a wide range of vehicle types are accounted for within the model and the addition of the CO\(_2\) calculation was new for version 6.2 [US EPA, 2003].

A different use of agent based modelling is seen in a study by Shafiei et al (2012b) where an agent-based modelling approach is applied to explore uptake of Electric Vehicles (EVs). They used a Multinomial Logit (MNL) model to characterise car purchase decisions and analyse the willingness of agents to consider an EV over a conventional Internal Combustion Engine. The model also allowed a study of the extent to which exposure to information affects willingness to consider the alternatives. The data requirements of this approach lie in the parameters for characterisation of the agents. The model outputs information about the market share of EVs over the scenario period. Shafiei et al (2012b) demonstrate the application of agent-based modelling approaches to the transport sector and their ability to capture behavioural dynamics in the market. While this modelling approach does not output direct emission calculations from road transport, it provides a useful
insight into the fuel mix of the future vehicle fleet. This is useful in understanding future emissions from road transport and the extent to which emission reduction targets might be met by uptake of alternatively fuelled vehicles.

System Dynamics

Fishwick (2007) describes a dynamic model as one which captures the way a system changes in time and this type of approach has been used across a range of disciplines in the physical and social sciences. System dynamics models are based around causal loop diagrams (CLD) that simulate and analyse relationships and mathematical modelling of stocks and flows (Stemman, 2000). They incorporate both qualitative and quantitative techniques and analysis, making them highly versatile tools (Pfaffenbichler et al., 2010). Shafiei et al (2012a) describe system dynamics modelling of complex systems as a “top-down approach that looks at the process of market developments as a whole and facilitates understanding the interactions of many stakeholders…” (Shafiei et al., 2012a, p.45).

CLD are the main technique used within system dynamics modelling to explore the qualitative relationships between different aspects of the system. They work by identifying entities, which are the aspects of the system that can affect other aspects and can be affected by others, and they must represent an unspecified quantity. There are also exogenous factors within system dynamics that are pre-defined constants interacting with the identified entities. These entities are then connected by directional causal links, which form the loops that are reinforcing or balancing, representing positive or negative feedbacks within the system, see Pfaffenbichler, (2011) for more details. System dynamics has been applied in transport because the feedback and linkages these models are able to capture are useful for identifying interactions within the transport system. Shepherd (2014) provides a review of the ways that system dynamics modelling has been used to capture the transport system. Figure 3 shows the environment sub-module in ASTRA which is responsible for the calculation of CO₂ emissions. The vehicle fleet in the model is driven by socio-economic factors and this is coupled with the transport sub-module to calculate fuel consumption which is used to calculate the in use emissions. Upstream embodied emissions in the fuel production and vehicle production are also accounted for in calculating the total emissions from the activity in the model (Rothengatter et al., 2000).
The calculation of CO$_2$ emissions in MARS is undertaken using the equation given below (Pfaffenbichler, 2003).

$$CO_2^{e_{ij}^{PC}}(t) = a_2 \times \left(V_{ij}^{PC}(t)\right)^2 + a_1 \times V_{ij}^{PC}(t) + a_0$$

Equation 3

Where
Specific carbon dioxide emissions of the mode car for a trip from $i$ to $j$ in year $t$ (g/Vh-km)

Average speed for a car trip from $i$ to $j$ in year $t$ depending on the applied policy instrument vector output of MARS (km/h)

Parameters, source MEET project [Samaras and Ntziachristos, 1998]

Techno-economic models

Techno-economic models include a range of tools that look at the relationships between technology, the economy and the wider impacts of this and can also be combined with socioeconomic data [Anable et al., 2012]. E3 models are a subset of this category and capture the interactions between energy, economy and the environment. These models tend to be macro-scale models, sometimes looking at transport only as a sub-sector of wider economic activity [Schafer, 2012]. They have the ability to deliver insight on sector level emissions and generally provide aggregate outputs. As with many macro-scale models, the detail captured in traffic...
models or behavioural approaches are often overlooked but there are examples of attempts to integrate some of this detail into these approaches. Figure 4 shows the process for calculating emissions in the ICCT Roadmap model, a techno-economic model of the transport system. It uses socio-economic indicators coupled with vehicle and fuel technology projections to estimate the emissions from road transport into the future, through the process shown in Figure 5 below.

Figure 4 - Simplified emission calculation in Roadmap [adapted from International Council for Clean Transportation, 2012]

Figure 5, shows an example of the outputs from the Roadmap model. On the right hand side of the figure is a base case scenario for global WTW CO$_{2e}$ emissions to 2030. The left had side has been subjected to a policy trajectory by adjusting the parameters shows in figure 5. Outputs can also be delivered for specific modes or regions. This demonstrates the value that this kind of model can have in understanding macro patterns of emissions, future projections and modal share, as
well as providing insight into regional responsibilities, which can be used and the international policy making level.

Figure 5 - Outputs from Roadmap for global WTW transport CO$_2$e emissions to 2050

IMACLIM-R is a hybrid-dynamic equilibrium model that explores emissions from the transport sector for 2001-2100 (Waisman et al., 2013). The model characterises transport emissions as based on four factors:

- Technological
  - Intensity of fuels
  - Energy intensity of mobility
- Behavioural
  - Modal structure
  - Volume of mobility

The model captures non-energy and non-price drivers of transport dynamics as a bridge between top-down and bottom-up modelling approaches. The model includes specific features for rebound effects, induced traffic, modal breakdowns and impacts of land-use decisions (Waisman et al., 2013).

An example of E3 modelling of transport is given within the US Energy Information Administration’s (EIA) World Energy Projection System Plus (WEPS+) (US Energy Information Administration (EIA), 2011). Within WEPS+, the Transportation Sector Model (ITRAN) captures the transport sector. The model is a large scale model that provides aggregate sector emission projections. These outputs are based on socio-
economic activity, GDP and population figures in a macroeconomic model, and fuel
prices from the refinery/electricity models. Outputs from the model include service
demand in passenger miles by sub-mode, fuel and region, and service intensity in
passenger miles per Btu by sub-mode, fuel and region. [US Energy Information
Administration (EIA), 2011].

These models are providing aggregate emission projections for the transport system
which are at a large spatial scale. Unlike the micro-scale approaches, individuals or
vehicle dynamics are not apparent but the insight can be useful for national and
international policy and decision making. In particular, demonstrating the magnitude
and geographical distribution of emissions can proportion responsibility for those
emissions and the need for delivery of emission reduction measures.

Integrated Assessment Models

Integrated Assessment Models (IAM) can be used to model the interactions between
the economy and the environment across the entire system in order to understand
long term changes and processes at the macro-scale. They are able to capture the
transport sector as a component of economic activity and some models allow the
running of sectoral modules independently. They are similar in scale and scope to
techno-economic models, however they are designed to further explore the
environmental changes associated with the economy.

The Global Change Assessment Model (GCAM) is an open licensed IAM [Kim et al.,
2012]. It is a partial equilibrium model which allows the separate areas of the
economy to be modelled independently. It is a large scale model that produces
aggregate emission projections, but the transport module can deliver insight on
future technological pathways and mode shares for the global regions [Kim et al.,
2012]. The equations below show how emissions from transport are calculated in

\[ D_{rt} = \alpha (Y_{r,t})^\alpha (P_{r,t})^\beta (N_{r,t}) \]  

Equation 4

Demand for passenger transport in region r and time period t, D_{rt}, is calculated using
the following:

\[ D_{rt} = \sigma (Y_{r,t})^\alpha (P_{r,t})^\beta (N_{r,t}) \]

Equation 4

Where:
\[ \sigma = \text{base year (2005) calibration parameter} \]
\[ Y = \text{GDP per capita} \]
\[ P = \text{total service price aggregated across all modes} \]
\[ N = \text{population} \]
\( \alpha = \text{income elasticity} \)

\( \beta = \text{price elasticity} \)

Total cost of any passenger mode \((P_i)\) in region \(r\) and time period \(t\) is calculated using the following:

\[
P_{i,r,t} = \frac{(FP_{i,r,t})(I_{i,r,t}) + NFP_{i,r,t}}{LF_{i,r,t}} + \frac{W_{r,t}}{S_{i,r,t}}
\]

Equation 5

Where:

\( FP = \) fuel price

\( I = \) vehicle fuel intensity

\( NFP = \) vehicle non-fuel price

\( LF = \) load factor

\( W = \) wage rate

\( S = \) vehicle speed

Market share of a given mode \(S_i\) is calculated as follows:

\[
S_{i,r,t} = \frac{(SW_{i,r})(P_{i,r,t})^\lambda}{\sum_i^n(SW_{i,r})(P_{i,r,t})^\lambda}
\]

Equation 6

Where:

\( SW = \) share weight

\( P_i = \) cost of transport service from above

\( \lambda = \) cost distribution parameter

\( n = \) number of modes

These equations demonstrate how emissions from road transport are calculated within the IAM, GCAM. As is the case with techno-economic approaches, aggregate outputs concerning emissions from transport are produced from this type of model. Additional changes in technology, load factors and modal share can also be captured, and the impacts on emissions from the transport sector be analysed. These aggregate outputs are useful for policy and decision making and the national and international levels.

**Combining modelling approaches**

Combining multiple modelling techniques and approaches can deliver a new perspective of the transport sector or generate new understanding of the dimensions of the transport system (e.g. Shafiei et al., 2012a; Köhler et al., 2009; Samaras et al., 2012). As Shafiei et al (2012a) suggest, "The real world problems in the
transportation system, however, do not match up well with a single modelling approach’ (p.43).

Examples of combining models to create new approaches for understanding the transport system include Kohler et al (2009) and Shafiei et al (2012a) who combined agent based and system dynamics techniques to explore transitions for sustainable mobility. Combining these two approaches allowed the exploration of behavioural interactions amongst heterogeneous agents in the agent based modelling study components. The agent based modelling components are then interfaced with a systems dynamics tool to explore how these behaviours play out in interactions and relationships in a future transition to sustainable mobility. As a further example of model combination, Samaras et al (2012) combined a traffic network model with a model of ICT applications in transport to look at the emissions implications arising from the integration of ICT in transport. This combination of multiple tools allowed the exploration of emissions from new ICT enabled transport schemes that was otherwise challenging within the scope of a single modelling approach.

The EU GHG-TransPoRD project combined a series of modelling approaches to look at regional emissions from transport and understand trajectories for emission reductions in the transport sector (Schade and Krail, 2012; Fiorello et al., 2012). The research integrates three regional models including ASTRA, POLES (Prospective Outlook for the Long Term Energy System), TREMOVE and the metropolitan scale MARS model to deliver greater understanding of potential emissions of greenhouse gases at the EU scale towards 2050 (Fiorello et al., 2012). This combining of multiple approaches, including system dynamics modelling, can deliver insight into the complex interconnections of the international transport sector at this scale and meet the challenges of long term forecasting of emission reductions.

The use of integrated models within a single research study allow some of the short comings of individual modelling techniques to be overcome. For example, combining focused behavioural studies with broader models such as traffic network models or system dynamics tools allow the understanding of individuals’ mobility behaviour to be placed in the wider context. This supports delivery of outputs that can then be utilised by planners and decision makers in policy development, although the integration of multiple modelling approaches can result in models which are more complex and may be challenging to non-expert users.

**Summary**

The previous sections have presented a range of approaches to modelling the transport system, and in particular the emissions of CO₂ from that system.
Microsimulation, behavioural modelling and agent based models all represent transport networks at the micro scale, characterising network level interactions and the decisions that generate patterns of mobility. They are predominantly of value for local level decision making as they can deliver insight at the geographical and temporal scales required for these decisions. These models vary in the level of detail captured about the individual, for example, microsimulation models capture the emissions from individual vehicles, but the dynamics of behaviour and motivations behind activities are not captured, whereas the agent based and behavioural models capture greater details about these factors. The calculation of CO$_2$ emissions within these models varies, from applying an emission factor to the sum of travel demand generated to the more detailed calculations undertaken in VISSIM, with the add-on EnViVer Pro, which accounts for varying speed-time profiles and other factors that might affect the emissions [Smit et al., 2006]. The applicability of an individual approach depends on the question being asked, with greater disaggregation generating more data, and perhaps the simplest approach for the questions is necessary, and often policy questions only require high level aggregate emission estimates to understand the impacts of changes.

Techno-economic models and IAMs, however, allow insight at the scale of the regional or global economy, which can provide useful decision making at the national and international policy making levels [Paltsev et al., 2005]. The quantification of emissions of CO$_2$ from road transport is necessary at both local and global level and across the geographic and temporal scales in order that decisions can be effectively made that will mitigate emissions of CO$_2$ and avoid dangerous climate change. There are cases where the models are taken out of their conventional scales. For example, the EU TransTools model uses traffic network modelling principles to represent trips in Europe, providing a large volume of detail across a wide spatial area [Brun and Hansen, 2009]. However, the general trend in transport modelling is for high levels of disaggregation when exploring a smaller spatial scale, with increasing computational demand, e.g., in microsimulation tools or behavioural type studies [Wegener, 2011]. Figure 1 provided a graphical demonstration of the temporal and geographical spread of the models that have been reviewed in this paper, showing what the different approaches.

System dynamics modelling in transport is positioned between the small scale microsimulation studies and the large scale techno-economic models. For example MARS examines dynamics in the transport system and the metropolitan scale whilst ASTRA encompasses the European transport network. As suggested by Figure 1, this provides a set of tools that can project future transport systems at a scale which
sits between micro scale approaches and the large techno-economic models. Additional advantages of utilising the system dynamics approach is the capturing of feedback loops through the CLD diagrams which can capture the dynamics of multiple stakeholders and policy outcomes in a complex system [Shepherd, 2014]. Particular areas of strength in system dynamics modelling of transport lie in the use of tools to capture interactions between land-use and transport, e.g. MARS, and model of the markets for alternative vehicles, e.g. Shepherd et al [2012]. In terms of emissions, this can deliver insight in to the volume of emissions from the transport system under analysis, and the modelling of future vehicle markets yields insight on the potential for reduction in emissions of CO$_2$ from the vehicle fleet through uptake of alternative vehicles.

It is clear that whatever system scale the study requires, modelling tools are available that can provide insight to the transport challenges and the impacts of potential policies and solutions. In addition, the combining of multiple modelling techniques can deliver additional innovative approaches to modelling transport [Shafiei et al., 2012a].

Conclusions

Three research questions have underpinned this review of how transport models estimate the emissions generated by the transport system. The first was concerned with identifying the primary techniques and tools used to model transport systems. A classification of five main types has resulted, i.e. microsimulation, behavioural, agent based, techno-economic and IAMs. The rationale behind the classification was the scope addressed by the model i.e. in terms of geographic, temporal and aggregation scale. In addition, the potential to interface and integrate models has been described.

The second research question concerned the methods used by the models to capture emissions from transport. This aspect of the research is synthesised in column 5 of Table 1. Essentially each modelling approach implies some assumptions ranging from driver behaviour, vehicle market share, fuel consumption converting to CO2 or to aggregate demand for transport on a broad scale. The efficacy of each type of model in capturing emissions overall rests on two broad sets of assumptions. The first assumptions are those underlying the way in which the model represents the transport system. The second assumptions underpin the detailed algorithms and emissions factors that are then used to produce emissions estimates. Whilst the first set are broadly understood and well reported, the second set of assumptions is often provided in more detailed manuals and aimed at an expert user. The overall efficacy
of the model in each case will clearly rest on the appropriateness of both sets of assumptions to the context in which emissions are to be estimated.

The final research question was concerned with reflecting on the advantages and disadvantages of the different approaches. From the summary provided for each type of model, it is clear that each has advantages when used in the context (geographic, level of aggregation etc.) for which it was designed and disadvantages occur when their use is extended beyond that original context. The practice of interfacing or integrating models has arisen, at least in part, from a need to fill some of the gaps in modelling capability that are still apparent in estimating emissions at particular geographic or temporal scales.

Clearly the availability of different modelling tools makes the selection of the correct tool for the specific analysis important. Overall, different modelling perspectives and paradigms create a rich and diverse series of tools and techniques that can capture the transport system and resulting emissions. These form a highly valuable resource in a field where technology, economics, behaviour and other factors are all key to understanding the dynamics and the resulting emissions from the sector. In addition, combining different modelling approaches can deliver further insights and understanding of the emissions from transport, as often a single approach cannot fully capture the dynamics of such a complex system.

This review has outlined the ways in which transport models can capture emissions from the sector, the different approaches and techniques, and discussed the value of these. By focusing on the quantification of emissions in transport modelling this paper contributes to the substantive body of literature on transport modelling whilst providing a novel perspective, with potential value to the academic and practitioner communities.

References


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