



Electric Vehicles Insights

Prepared by ENERGEIA for the Australian
Energy Market Operator's 2017 Electricity
Forecast Insights

September 2017

Executive Summary

The AEMO Electricity Forecasting Insights report provides electric vehicle consumption forecasts to 2050 for the National Electricity Market (NEM), and for each NEM region.

While electric vehicle uptake in Australia is still very low (approximately 0.1 of annual vehicle sales¹), the combined impact of price declines in battery technology, the increasing introduction of new EV models into the market and both government and industry support is expected to drive rising uptake over the next 20 years.

AEMO has commissioned Energeia to prepare an Electric Vehicles Insights paper and to adopt the scenario assumptions of AEMO's recently published 2017 Electricity Forecasting Insights report as the basis for an impact assessment of the introduction of electric vehicles on Australia's electricity supply system.

Over the course of 2017, AEMO will monitor feedback on this report, and continue a work-program to enable the inclusion of electric vehicles in AEMO's major Forecasting and Planning publications in 2018.

Scope and Approach

The Electric Vehicle Insights paper provides Energeia's scenario based forecasts of EV uptake for each region of the NEM and the corresponding impact on annual electricity consumption, maximum and minimum demand because of charging of EVs from the grid.

Energeia has used its fourth generation EV forecasting model, updated to align with AEMO's scenario based assumptions regarding electricity prices as well as market and policy settings, to derive the results.

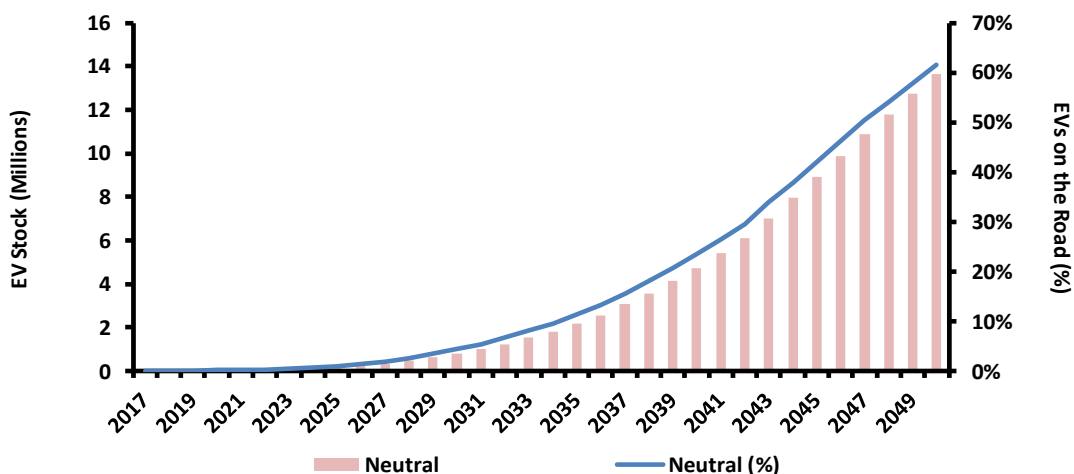
Results

The following sections reports on the results of the middle scenario. Results for the high and low scenarios are provided in Section 4.

EV Uptake

EV sales within the National Electricity Market (NEM) are forecast to reach 431,000 vehicles per annum by 2036 or 36.5 of new vehicle sales under AEMO's neutral sensitivity, increasing to 1.58 million or 90% of new vehicle sales by 2050. As a result, total EVs on the road are forecast to reach over 2.56 million or 13.2% of total new sale vehicles by 2036, moving to 13.63 million or 61.5% of all new vehicles sales by 2050, as shown in Figure 1.

Figure 1 – EV Uptake (NEM, Neutral)



Source: Energeia

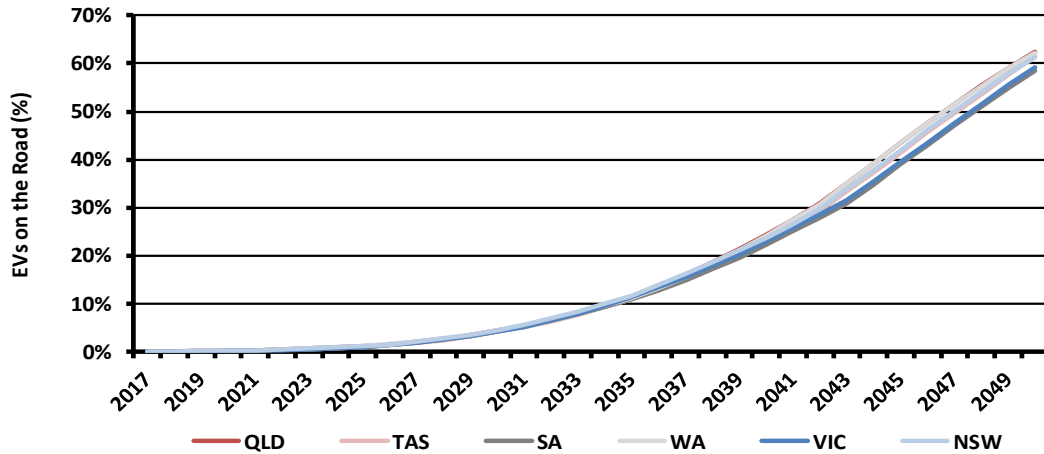
¹ ClimateWorks Australia; The state of electric vehicles in Australia; June 2017

The difference between the percentage of EVs sold and the percentage EVs on the road is due to the average length of time before Australians retire their vehicles, which currently stands at just over 22 years.²

The main drivers of increasing adoption rates over time are rising EV model availability, elimination of purchase premiums and falling battery storage costs.

The variation in vehicle stock forecasts shown in Figure 2 is due to the relative differential between petrol prices and electricity prices in each state, with QLD having the greatest differential due to its relatively low cost controlled load tariffs.

Figure 2 – EV Uptake by Region (Neutral)

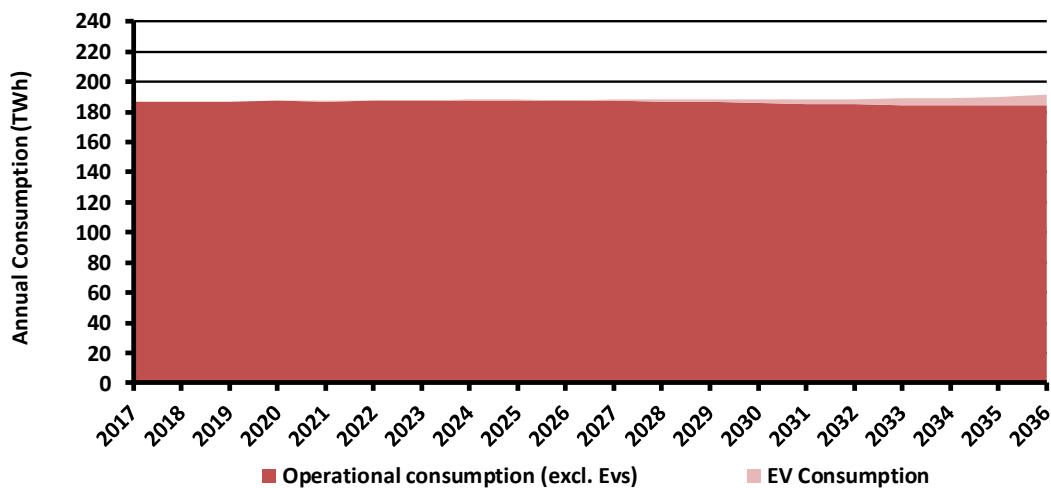


Source: Energeia

EV Consumption

Energeia's consumption forecast under the middle scenario is for EVs to consume around 7.02 TWh of grid electricity per year by 2036, increasing total consumption by around 3.8% compared to AEMO's forecasts for primary load under the neutral sensitivity as shown in Figure 3 below.

Figure 3 – EV Electricity Consumption Compared to Insights Forecast (NEM Operational, Neutral)



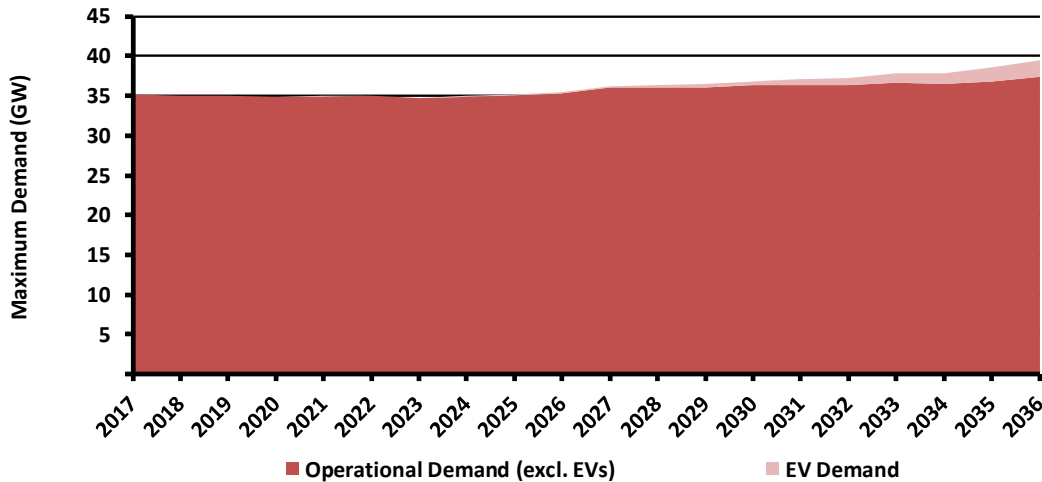
Source: Energeia

² ABS 9208.0 - Survey of Motor Vehicle Use, Australia, 12 months ended 30 June 2016

EV Maximum Demand

Energeia forecast of non-coincident aggregated EV maximum demand by region is shown in Figure 4 below. Energeia’s modelling optimises flexible electric vehicle load to avoid contributing to network and system load. This means that flexible electric vehicle peak demand is likely to occur after the typical evening peak.

Figure 4 – EV Maximum Demand by Region (NEM Operational, Neutral)



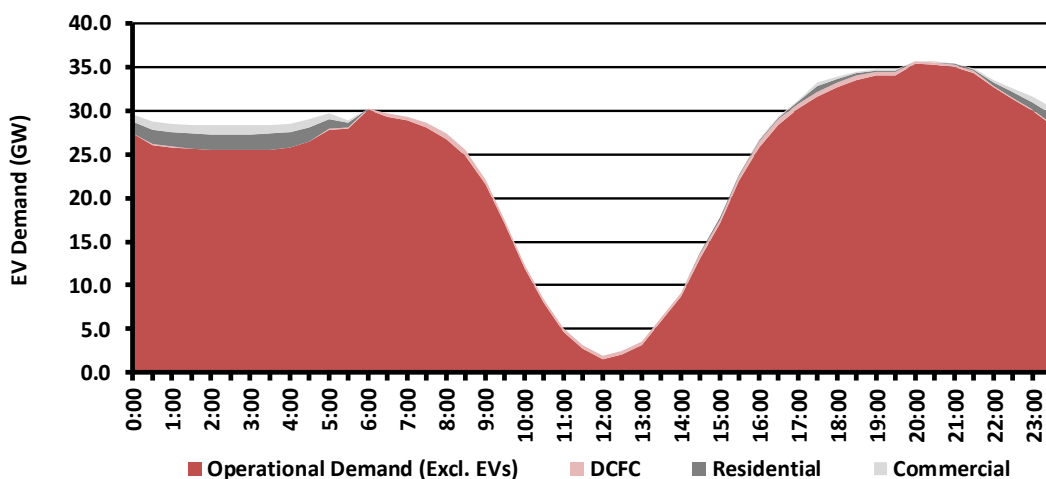
Source: Energeia

Electric vehicle charging optimisation applies to home charging of passenger vehicles and workplace charging of commercial vehicles. Fast charging is not assumed, via the charge control system, to manage demand, as the value of the service is the speed of the recharge.

Impact on Maximum and Minimum Regional Demand

Despite the increase in consumption and load from electric vehicles, Energeia’s modelling shows their peak demand impacts as being limited to fast charging sub-segment only. Flexible home and workplace charging load is able to be orchestrated to avoid contributing to network or system peak demand, as shown in Figure 5 below.

Figure 5 - EV Charging Load by Charge Type (NEM, 2036)



With respect to the impact of electric vehicle load on minimum demand, Energeia’s modelling shows mainly fast charging increasing minimum demand, even as it shifts to the middle of the day due to solar PV. Home charging of non-commuter vehicles is also expected to increase minimum demand. Workplace charging of commercial vehicles is not expected to impact minimum demand once it shifts to the middle of the day.

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1 Introduction

1.1 Background

The Electricity Forecasting Insights report provides electricity consumption forecasts over a 20-year forecast period for the National Electricity Market (NEM), and for each NEM region. In 2015, the NEFR considered the impact of uptake of electric vehicles (EVs) for the first time via the Emerging Technologies Paper accompanying the 2015 NEFR³. 2016 was the first year that an Electricity Forecasting Insights report was prepared.⁴

AEMO has commissioned Energeia to update the Electric Vehicles Insights report in 2017 to take account of key developments in the electric vehicle market since the 2016 report was published, including unforeseen changes in the availability of the most popular EV models including the Mitsubishi Outlander and the Nissan Leaf. This year, the scope has also been expanded to include commercial and DC fast charging vehicles.

AEMO uses the Electricity Forecasting Insights report analysis as the starting point for monitoring the emergence and use of electric vehicles in Australia, as well as an input to its modelling integration studies and the 2017 Electricity Forecasting Insights.

AEMO will monitor feedback on this report, and will continue a work-program to develop this analysis to enable the inclusion of electric vehicles in AEMO's major Forecasting and Planning publications in 2017. A further objective is to provide recommendations as to how the EV forecasts may be better integrated into the Electricity Forecasting Insights report in future years to continually improve forecasting accuracy.

1.2 Objectives

The primary objective of this Electric Vehicles Insights paper is to reduce the potential forecasting uncertainty within the main AEMO forecasts with respect to EV uptake by providing a specialist forecast of EV uptake and charging demand over time.

Specifically, the paper provides forecasts of EV uptake for each region of the NEM and the corresponding impact on annual electricity consumption and maximum and minimum demand due to charging of EVs from the grid.

1.3 Scope and Approach

The EV forecasts consider impacts from EVs taken up within the passenger vehicle and commercial sectors, excluding (heavy) articulated trucks and speciality vehicles such as bucket trucks. The passenger sector includes passenger cars and sport utility vehicles. The commercial sector includes light commercial (vans and trucks), buses, and rigid vehicles. EV forecasts include both battery electric vehicles (BEVs) and plug-in hybrid vehicles

Battery Electric Vehicle (BEV) – Powered only by energy stored in batteries with batteries charged by plugging into the grid.

Internal Combustion Engine Vehicle (ICE) – Represents most private vehicles, powered by a standard internal combustion engine using petrol, diesel or gas.

Hybrid Electric Vehicle (HEV) – Combines both an ICE with an electric engine. The electrical energy is stored in a battery that is charged by the internal combustion engine. Battery capacity is generally limited. Vehicle propulsion is a mix of the ICE and electric engine, but is predominantly powered by the ICE. Does not take energy from the electricity grid.

Plug-in Hybrid Electric Vehicle (PHEV) – Combines both an ICE with an electric engine. Electrical energy is stored in batteries by plugging into the grid. Vehicle propulsion is a mix of the ICE and electric engine, but is predominantly powered by the electric engine. The ICE is used to extend driving range beyond battery capacity for longer distances and to recharge the battery itself.

³ Emerging Technologies Information Paper, National Electricity Forecasting Report Published: June 2015

⁴ AEMO Electricity Forecasting Insights 2016

(PHEVs) to the extent that they utilise the grid for charging. The forecasts exclude hybrid electric vehicles (HEVs) which do not charge from the grid.

Energeia has used its fourth generation EV forecasting model (described further in Section 2), updated to align with AEMO's assumptions regarding electricity prices as well as market and policy settings, to derive the results. Further specific EV assumptions were set in conjunction with AEMO as described in Section 3.

1.4 Limitations

The EV forecasts contained throughout this paper are independent of the base AEMO forecasts. That is, there is no feedback loop between the forecasted EV uptake and the corresponding response from networks, retailers or the wholesale market.

Further, there are a range of future possibilities as to how EV loads will be priced and how the EV market will integrate with the electricity market and it is foreseeable that tariff products could evolve to encourage increased charging of EVs during solar generation times. This analysis assumes initial EV tariffs for home and workplace charging reflect controlled load tariffs, which will be orchestrated to ensure they minimise peak demand impacts.

There is also likely to be some degree of interaction between solar PV, stationary battery storage and EVs at residential and workplace premises. While AEMO has separately undertaken solar PV and battery storage forecasts, these have not been integrated with the EV forecasts in this paper.

The forecasts cover EV charging loads at home for passenger vehicles and at business locations for commercial vehicles. They also cover DC fast charging loads to serve long-haul commercial vehicles like busses and articulated trucks, as well as passenger vehicles that are expected to not have access to off-street parking based home chargers, such as second and third vehicles in a given household.

The household transport model upon which the EV forecast model relies are derived from the Queensland Household Travel Survey and Victorian Managing Tariff Congestion Report. That is, while the model reflects different average driving distances between states, it assumes that travel patterns (origins, destinations, arrival times and departure times) in all regions of Australia are consistent with those of Queensland drivers for passenger vehicles with access to private parking, while travel patterns for commercial EVs and vehicles without access to private parking are consistent with drivers in Victoria.

The EV uptake model is driven in part by the financial return on investment to vehicles owners based on the EV vehicle premium and reduced operational costs. The model does not consider any costs associated with any required upgrade to the household switch board and/or service, which could add considerable cost. However, this is not expected to be a material number of households based on anecdotal evidence from pilots, etc.

While all due care has been taken in the preparation of this paper, Energeia has relied upon stakeholder provided information as well as publicly available data and information. To the extent these reliances have been made, Energeia does not guarantee nor warrant the accuracy of this paper. Furthermore, neither Energeia nor its Directors or employees will accept liability for any losses related to this paper arising from these reliances.

The forecasts derived from Energeia's EV forecast model are supplied in good faith and reflect the knowledge, expertise and experience of the consultants involved. Energeia does not warrant the accuracy of the model nor accept any responsibility whatsoever for any loss occasioned by any person acting or refraining from action as a result of reliance on the model. The model and this report are for educational purposes only.

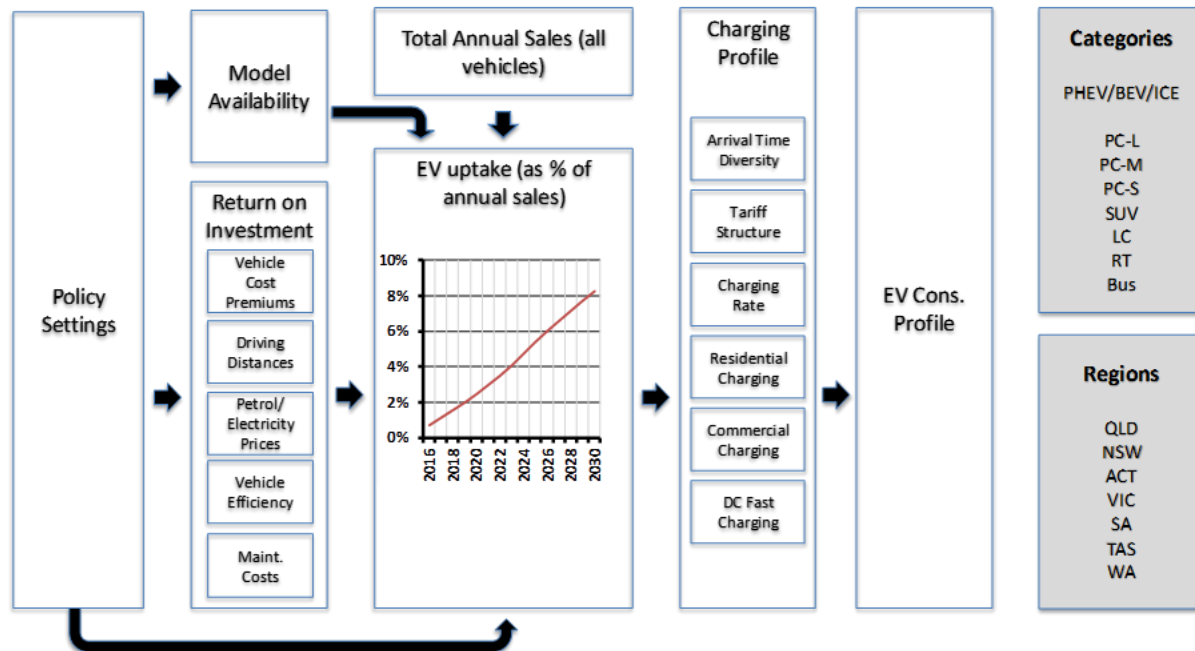
2 EV Forecasting Model Overview

The following section provides an overview of the Energeia's EV forecasting model. The model is part of Energeia's broader energy system model, but has been limited in this study to identify the (non-integrated) impacts of electric vehicles on the energy system. Detailed modelling assumptions are provided in Appendix A.

2.1 Overview

Energeia's EV forecasting model is comprised of two parts, EV uptake and EV charging as shown in Figure 6 below.

Figure 6 – Energeia EV Forecasting Model



Source: Energeia

The EV uptake module forecasts EV uptake for each category of vehicle using vehicle model availability and the vehicle owner's return on investment as inputs. The forecast is allocated on a pro-rata basis to each state based on the state's 2016 share of vehicles on the road based on ABS data.⁵ The EV charging module then applies a charging regime to each vehicle adopted based on its:

- charging type,
- arrival and departure time for home and workplace charging or transportation profile for DC fast charging,
- the number of kilometres travelled and
- grid load to optimise workplace and home charging.

The model considers 8 categories of vehicle types including:

- Vehicle class
 - Passenger Car Large (PC-L)
 - Passenger Car Medium (PC-M)
 - Passenger Car Small (PC-S)

⁵ ABS 9208.0 - Survey of Motor Vehicle Use, Australia, 12 months ended 30 June 2016

- Sport Utility Vehicle Medium (SUV-M)
- Sport Utility Vehicle Large (SUV-L)
- Light Commercial (LC)
- Rigid Truck (RT)
- Bus (B)

Each of these categories have specific characteristics which drive both uptake and charging, including:

- purchase premium,
- energy consumption per km, and
- battery size.

Fuel costs and average daily driving are based on state level factors.

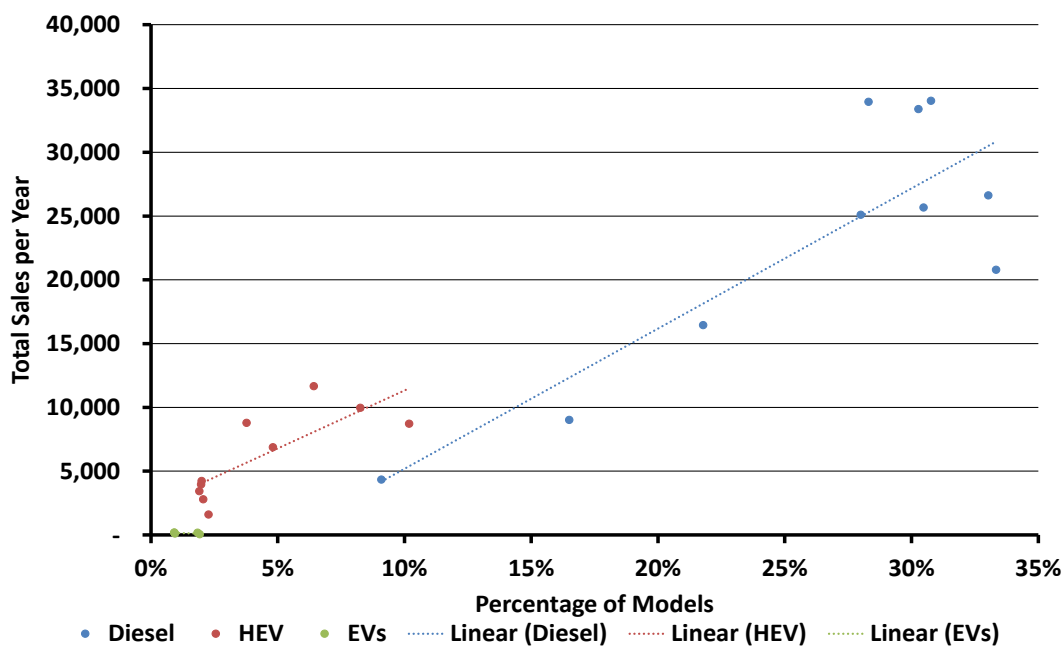
2.2 EV Uptake

EV uptake is determined by a two-parameter function that describes vehicle uptake over time based on:

- Model Availability: The percentage of models within a given vehicle class available in EV form
- Return on Investment: The first-year return to the vehicle owner investing in an EV in terms of reduced operational costs (fuel and costs) on the premium paid compared to a conventional ICE vehicle, net of any purchase incentives.

This functional form accordingly considers the supply side constraints (lack of model availability) as well as demand side drivers (reduced operational costs) in the vehicles owner’s decision to adopt. The function is derived from analysis of diesel vehicle and hybrid electric vehicle adoption patterns in Australia which showed uptake was best explained by a combination of these parameters. The historical relationship between vehicle uptake and model availability in the Australia market for alternative technologies is shown in Figure 7 below.

Figure 7 – Relationship between EV Uptake and Model Availability



Source: VFACTS, Energeia

Detailed assumptions driving the EV Uptake Model are provided in Appendix A.

2.3 EV Charging

The EV charging profile is determined by aggregating the charging profile of each electric vehicle adopted by charging type and state segment as follows:

- Whether the vehicle is assigned as L2 home charging, L2 commercial charging (charges at work or depot location), or Direct Current Fast Charging (DCFC) which is defined as the EV equivalent of a gas station (1MW station with 5 min charge time by 2036)
- The average expected daily travel distance by state, which determines the amount of charge to be supplied by day type via the charging profile (i.e. the area under the curve).
- Average expected arrival time by hour (drawn from a database of home and commercial arrivals times) which dictates when charging starts in the absence of any other tariff restrictions
- Average expected departure time by hour (drawn from a database of home and commercial departure times) which dictates when charging must cease in the absence of any other tariff restrictions
- The optimised home and workplace charging profiles that deliver the required level of charge within the arrival and departure times, without impacting on grid peak demand. The DC Fast Charging profile is based on transportation demand as no flexibility is assumed due to the nature of the service.

Detailed assumptions driving the EV charging profiles are provided in Appendix A.

3 Sensitivities

Three forecast sensitivities were modelled that represent the expected pathway for Australia across weak, neutral (considered the most likely), and strong economic, technical and consumer outlooks aligned with AEMO's broader forecast sensitivities. The results of the neutral sensitivity are reported on in the main body of this paper, however, forecast uptake, consumption and peak demand are reported for all sensitivities in Appendix B4.

3.1 AEMO Sensitivities

AEMO's 2017 Electricity Forecast Insights paper uses the terms "weak", "neutral", and "strong" throughout the report to identify the three sensitivities with the neutral sensitivity considered the most likely (i.e. the 'P50'). The weak and strong sensitivities are based on dynamics affecting the total energy consumption of households and businesses and are not necessarily a low and high outcome for the consumption of grid-supplied energy, but rather an internally consistent set of assumptions aligned to strong and weak economies, technology change and consumer sentiment. The key characteristics of these sensitivities of relevance to EVs are shown in Table 1.

Table 1 – 201 AEMO Insights Sensitivity Drivers

Driver	Weak Sensitivity	Neutral Sensitivity	Strong Sensitivity
Population Growth	ABS projection C	ABS projection B	ABS projection A
Economic Growth	Weak	Neutral	Strong
Electricity Network Charges, 5 Years	Current AER determinations, fixed after 5 years		
Electricity Retail Costs and Margin	Assume current margins throughout		
Oil Prices	UD30/bbl (BR) over 5 year glide path	UD60/bbl (BR) over 5 year glide path	UD90/bbl (BR) over 5 year glide path
Climate Policy	Assume Australia's Paris commitment is achieved		

3.2 EV Sensitivities

In addition to the AEMO Electricity Forecast Insight sensitivities, the EV Forecast Insights include the additional assumptions listed in Table 2.

Detailed assumptions underpinning the EV sensitivities are provided in Appendix A.

Table 2 – Additional EV Sensitivity Drivers

Driver	Weak Sensitivity	Neutral Sensitivity	Strong Sensitivity
EV Incentive	\$1,500	\$2,500	\$5,000
Year Incentive Applies	2023	2021	2019
EV Vehicle Parity	8	6	4
Tariff Settings (Home and Depot Charging)	Current controlled load tariffs (generally allowing overnight charging only)		
Tariff Settings (DCFC)	Vehicles without private parking available charge using DCFC charging stations		
Upper EV Limit	80%	90%	100%
Model Availability Ramp (Models/Yr)	2.5%	5%	7.5%
Vehicle Emission Standards	Commonwealth Government introduces international best practice emission standards (as fleet wide target) by 2023*	Commonwealth Government introduces international best practice emission standards (as fleet wide target) by 2021*	Commonwealth Government introduces international best practice emission standards (as fleet wide target) by 2019*

* A fleet wide standard has been assumed, rather than a minimum performance standard, as the most economically efficient means of achieving best practice greenhouse gas emission

4 Results

The results shown below describe forecast EV uptake over the period 2017 to 2050 and the corresponding contribution to energy consumption and both maximum and minimum demand at the NEM level and by state. The results are presented for the neutral sensitivity unless otherwise indicated.

Detailed forecasts by scenario and region are reported in Appendix B.

4.1 EV Uptake Forecasts

Section 4.1 presents uptake of EVs in terms of both annual sales and number of vehicles on the road (stock).

4.1.1 NEM

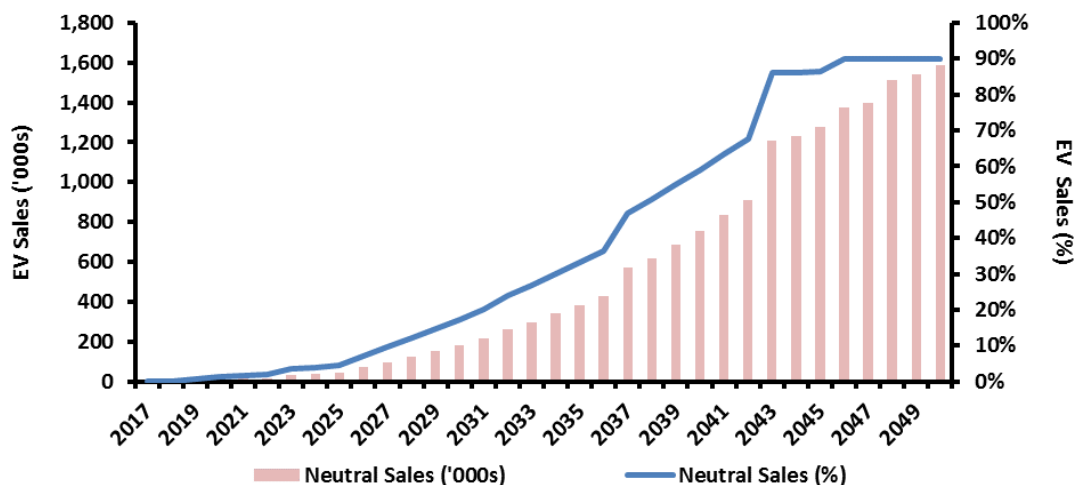
EV sales (both BEV and PHEV) are forecast to reach 431,000 vehicles per annum by 2036, increasing to 1.58 million annual new vehicle sales by 2050, or 36.5% and 90% of sales respectively as shown in Figure 8 – Annual EV Sales (Neutral) Figure 8 below.

Detailed EV modelling shows a relatively steady increase in EV sales to around 40% per annum by 2036 driven by:

- falling EV prices supported by falling battery prices,
- increased model availability by original equipment manufacturers (OEM), and
- an increasing differential between electricity and petrol prices.

Sales are forecast to see a step change in sales from 2036, when the first EVs segments begin to see two-year paybacks, reaching a market tipping point. From 2036 to 2042, annual sales growth is higher mainly due to falling battery costs. In 2043, there is another wave of EV segments reaching the 2-year payback threshold triggering a rapid market share increase up to the sensitivity limit. However, from this point, sales growth begins to taper off as the market reaches saturation.

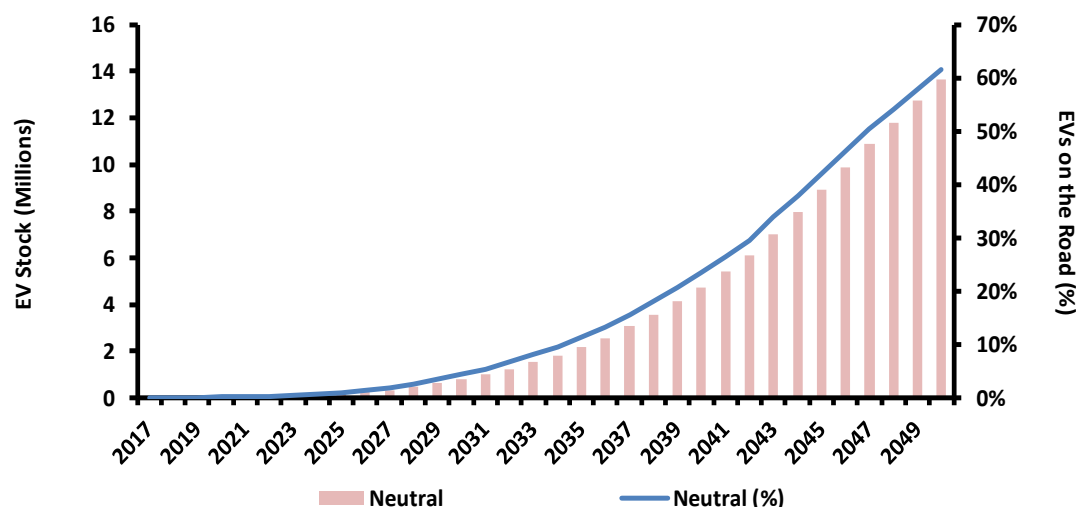
Figure 8 – Annual EV Sales (Neutral)



Source: Energeia

As a result, total vehicles on the road are forecast to reach 2.56 million by 2036 or 13.2% of vehicles. By 2050, EV uptake reaches 13.62 million vehicles, or 61.5% as shown in Figure 8.

Figure 9 – EVs on the Road (Neutral)



Source: Energeia

4.1.2 Regions

Uptake varies by region as shown in Table 3, predominantly due to market size, but also the relative differential between petrol prices and electricity prices experienced in each state in the early years, with QLD having the greatest differential due to its relatively low priced controlled load tariffs.

Table 3 – Annual EV Sales and Vehicles on the Road (Stock) by Region (Neutral)

Region	2017			2020			2030			2050		
	Yrly Sales (%)	Yrly Sales ('000s)	Stock ('000s)	Yrly Sales (%)	Yrly Sales ('000s)	Stock ('000s)	Yrly Sales (%)	Yrly Sales ('000s)	Stock ('000s)	Yrly Sales (%)	Yrly Sales ('000s)	Stock ('000s)
QLD	0.1%	0.1789	0.8852	1.4%	3.438	6.262	16.8%	46.0	199.0	89.8%	440	3750
NSW	0.1%	0.3320	1.5539	1.4%	5.765	10.573	17.7%	81.0	350.9	89.8%	672	5824
VIC	0.1%	0.1655	1.2073	1.4%	3.055	5.992	17.0%	44.3	189.6	89.7%	338	2879
SA	0.1%	0.0567	0.2819	1.3%	1.003	1.856	17.2%	14.4	60.6	89.8%	114	961
TAS	0.1%	0.0171	0.0741	1.4%	0.328	0.588	16.7%	4.3	18.8	89.8%	33	293
NEM	0.1%	0.7292	4.0439	1.4%	13.265	24.794	17.3%	182.8	791.5	89.8%	1585	13627
WA	0.1%	0.1211	0.4666	1.4%	2.224	3.946	17.0%	30.2	130.8	89.7%	308	2593

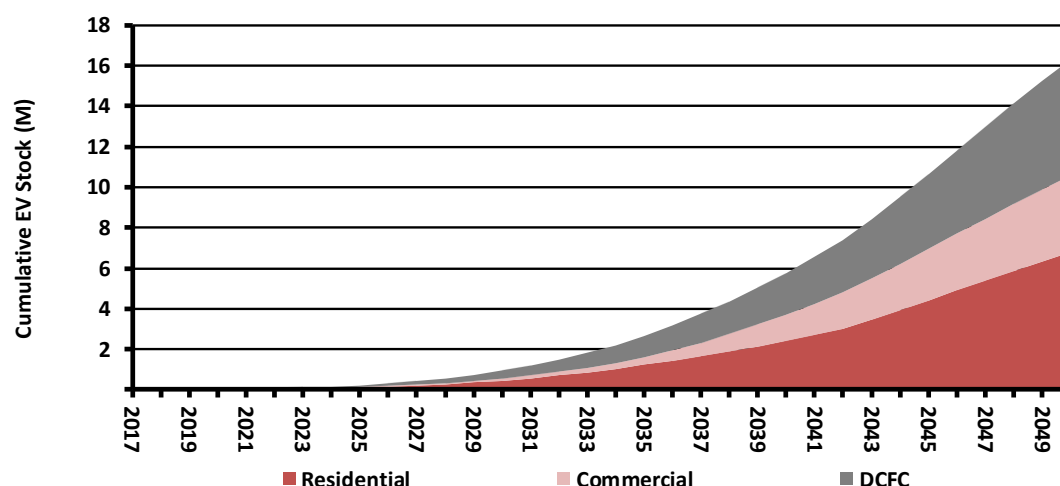
Source: Energeia

4.1.3 Charging Segment

EV Uptake by charging segment for the neutral scenario is shown in Figure 10.

The residential home charging segment is the largest market in terms of EV uptake, followed by the DCFC market and commercial markets respectively. By 2050, over 10 million of the 16 million EVs on the road will have private chargers available where they reach their destination, either home or at work. The remaining 6 million vehicles will use DCFC stations as a means to charge their vehicle.

Figure 10 - EVs on the Road by Charging Segment (Neutral)



4.1.4 Sensitivities

EV uptake forecasts vary significantly across sensitivity scenario as shown in Table 4 below for the NEM. Detailed results by region and sensitivity are presented in Appendix B.

In the strong sensitivity, EV sales initially increase at a faster rate than both the neutral or weak sensitivity due to the larger EV incentive, which applies earlier, and the earlier EV price parity. The strong sensitivity sales rate accelerates from 2024 due to a faster ramp-up of model availability driven in part by higher incentives. As a result, forecast EV stock under the strong sensitivity reaches 6.9 million vehicles by 2036, 2.7 times higher than the neutral sensitivity, and reaches 23.67 million vehicles by 2050, 1.7 times higher than the neutral scenario.

In the weak sensitivity, EV sales increase more slowly over time mostly due to a slower decline in EV price premiums and model availability. Under the weak sensitivity, the first EVs to reach the two year pay-back do so in 2042, five years later than the neutral scenario. As a result, forecast EV stock in the weak sensitivity reaches almost 1.27 million vehicles by 2036, 50% less than the neutral sensitivity. Looking further ahead, the EV stock under the weak sensitivity reaches 5.42 million vehicles by 2050, 60% less than the neutral scenario.

Table 4 – EV Uptake by Sensitivity (Neutral)

Sens.	2017			2020			2030			2050		
	Yrly Sales (%)	Yrly Sales ('000s)	Stock ('000s)	Yrly Sales (%)	Yrly Sales ('000s)	Stock ('000s)	Yrly Sales (%)	Yrly Sales ('000s)	Stock ('000s)	Yrly Sales (%)	Yrly Sales ('000s)	Stock ('000s)
Strong	0.1%	0.7454	4.0600	1.9%	18.990	32.368	40.8%	456.7	1839.8	99.8%	2510	23666
Neutral	0.1%	0.7292	4.0439	1.4%	13.265	24.794	17.3%	182.8	791.5	89.8%	1585	13627
Weak	0.1%	0.7135	4.0281	1.3%	12.068	23.130	9.5%	95.3	465.8	65.9%	855	5422

Source: Energeia

4.2 EV Consumption Forecasts

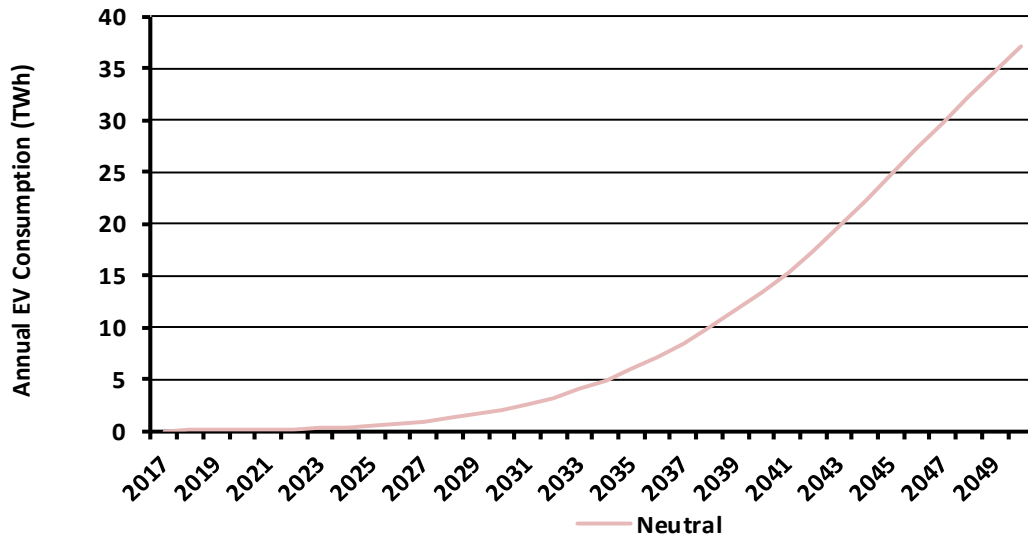
Section 4.2 presents the forecasts for grid electricity consumption from from Australia's EV stock charging and assesses the impact of these on the 2017 Insights forecasts prepared by AEMO. All of the forecasts are in terms of 'operational' requirements (including losses) as defined by AEMO⁶.

⁶ AEMO Electricity Forecasting Insights 2017

4.2.1 NEM

Energeia’s neutral sensitivity forecast sees EVs consuming around 7.02 TWh of electricity per year by 2036, and 37.08 TWh of electricity by 2050, as shown in Figure 12. The increase in EV consumption over time is directly related to the change in EV uptake as discussed in Section 4.1.

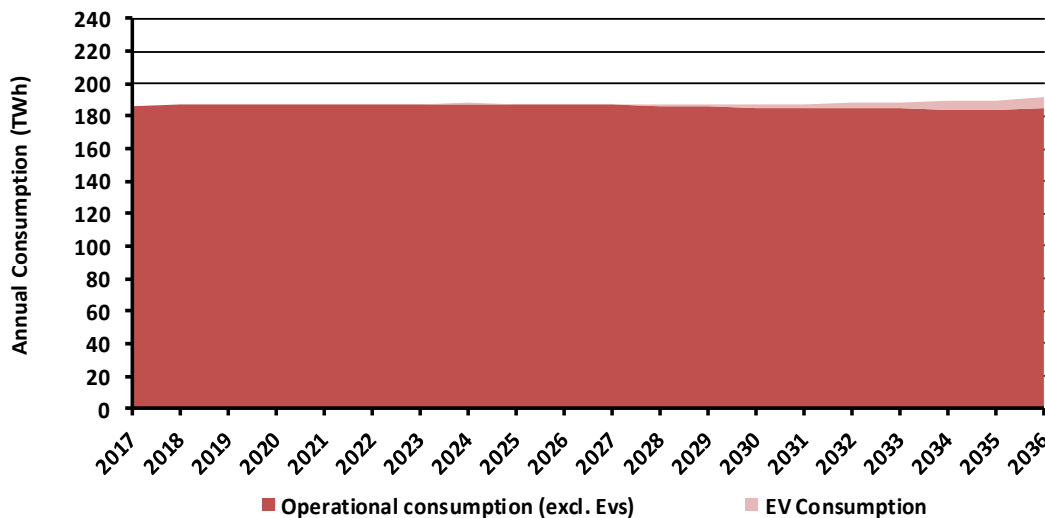
Figure 11 – EV Electricity Consumption (NEM Operational, Neutral)



Source: Energeia

The additional EV consumption is forecast to increase total consumption by around 3.81% compared to AEMO’s Insights forecasts for operational load in 2036 under the neutral sensitivity as shown in Figure 13 below.

Figure 12 – EV Electricity Consumption Compared to AEMO Forecast (NEM Operational, Neutral)



Source: Energeia

4.2.2 Regions

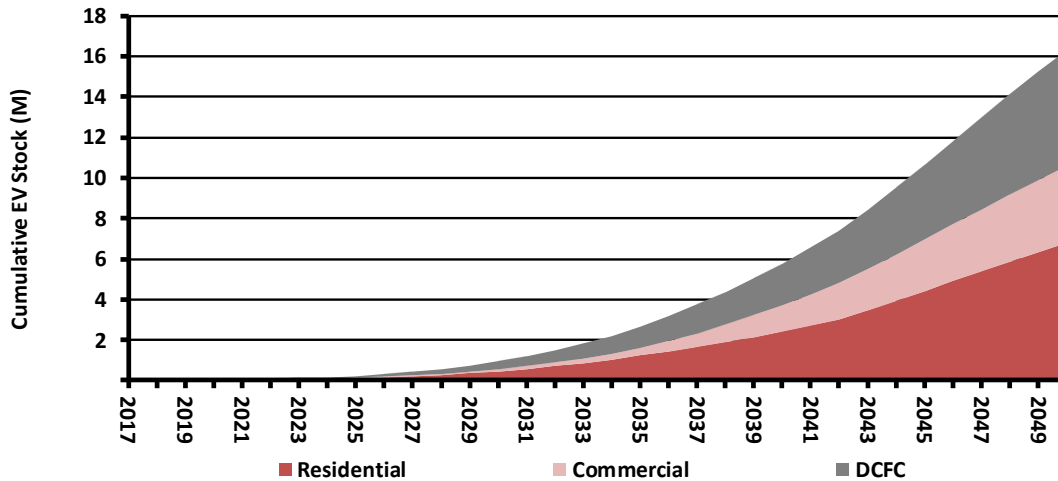
Electricity consumption by EVs is forecast to vary by region as shown in Figure 13 below.

The differences in regions are driven primarily by market size, with ACT and NSW having the largest market for new vehicles. The consumption aligns closely to EV uptake by region (as per Charging Segment

EV Uptake by charging segment for the neutral scenario is shown in Figure 10.

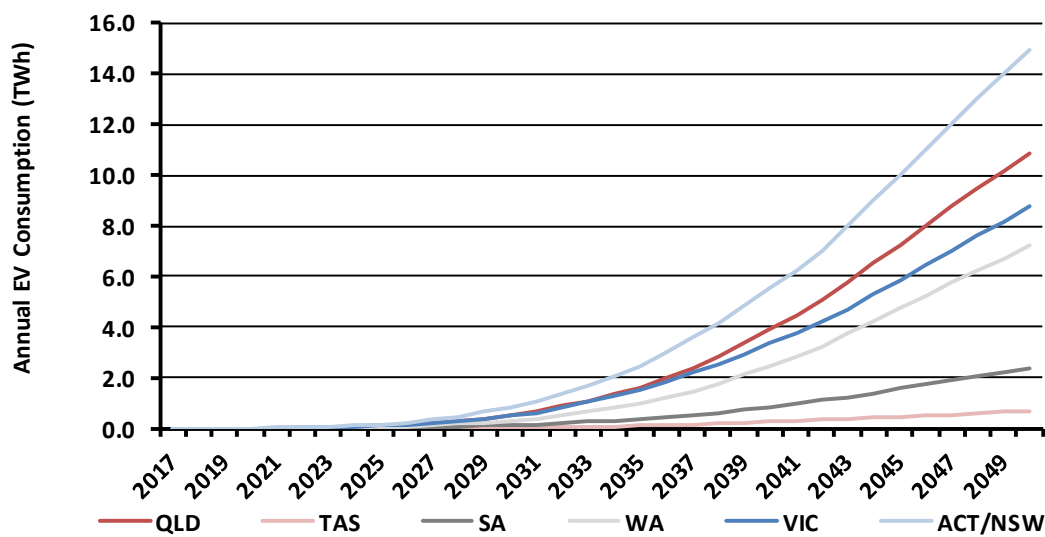
The residential home charging segment is the largest market in terms of EV uptake, followed by the DCFC market and commercial markets respectively. By 2050, over 10 million of the 16 million EVs on the road will have private chargers available where they reach their destination, either home or at work. The remaining 6 million vehicles will use DCFC stations as a means to charge their vehicle.

Figure 10 - EVs on the Road by Charging Segment (Neutral)



). Notwithstanding, EV consumption per vehicle does vary slightly by state due to the differences in average travel distances and tariff rates which in turn influence relative uptake of vehicle types (PHEV or BEV) and associated charging requirements.

Figure 13 – EV Electricity Consumption by Region (Neutral)



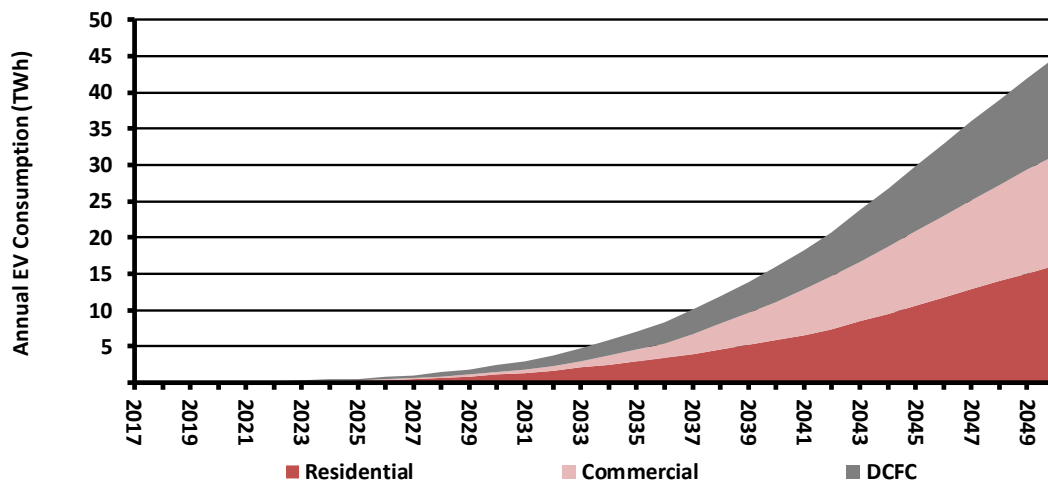
Source: Energeia

4.2.3 Charging Segment

Electricity consumption by charging segment for the neutral scenario is shown in Figure 13.

Unlike EV uptake, energy consumption between residential, commercial and DCFC is similar throughout the modelling period. Although the commercial market has around half the number of EVs on the road as the residential and DCFC market, additional distances travelled by commercial vehicles and the consumption requirements of these vehicles results in around double the energy needs of the other markets. The modelling shows that on average, residential and DCFC vehicles require 2.2MWh of electricity annually, while commercial vehicles require 4.4 MWh.

Figure 14 - EV Electricity Consumption by Charging Segment (Neutral)

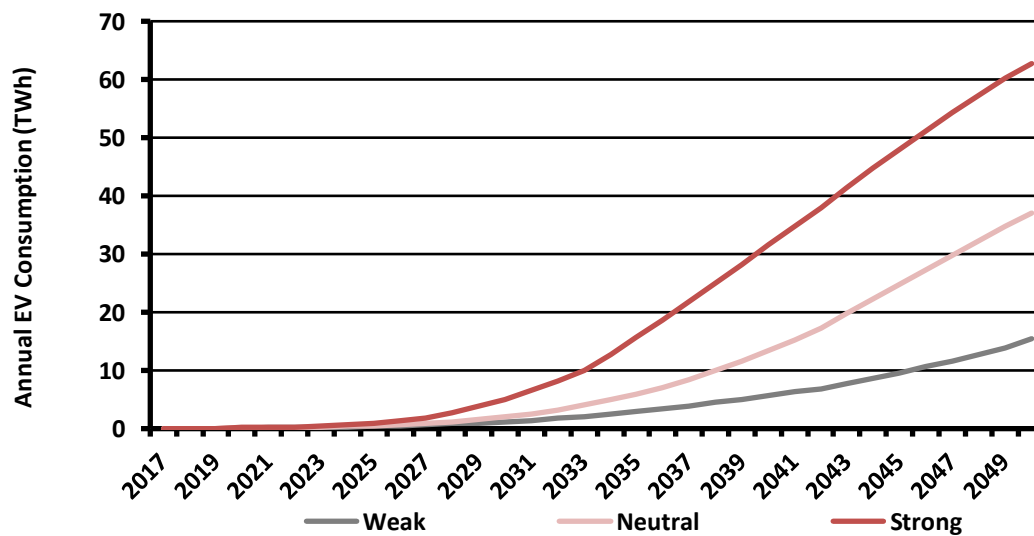


4.2.4 Sensitivities

EV electricity consumption forecasts vary significantly for the weak and strong sensitivities as shown in Figure 15 below for the NEM, as a result of the factors detailed in Section 4.1.2 on uptake.

Detailed results by region and sensitivity are presented in Appendix B.

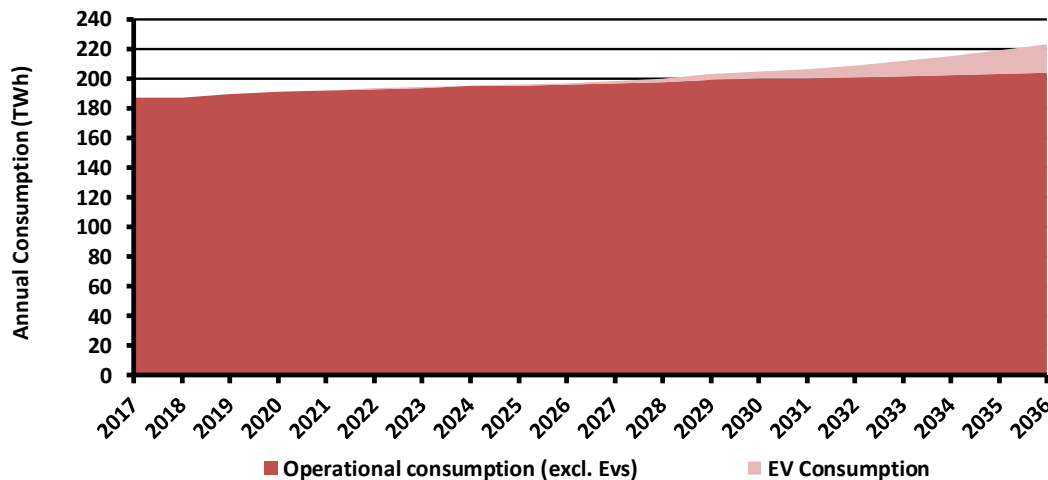
Figure 15 – EV Electricity Consumption by Sensitivity (NEM Operational)



Source: Energeia

Under the strong sensitivity, EV electricity consumption reaches 18.69 TWh per year and equates to around 9.16% of AEMO's Insights forecasts for operational load in 2036 as shown in Figure 16 below. This is around 266% above the neutral scenario, consistent with differences in uptake rates between the sensitivities.

Figure 16 – EV Electricity Consumption (NEM Operational, Strong)

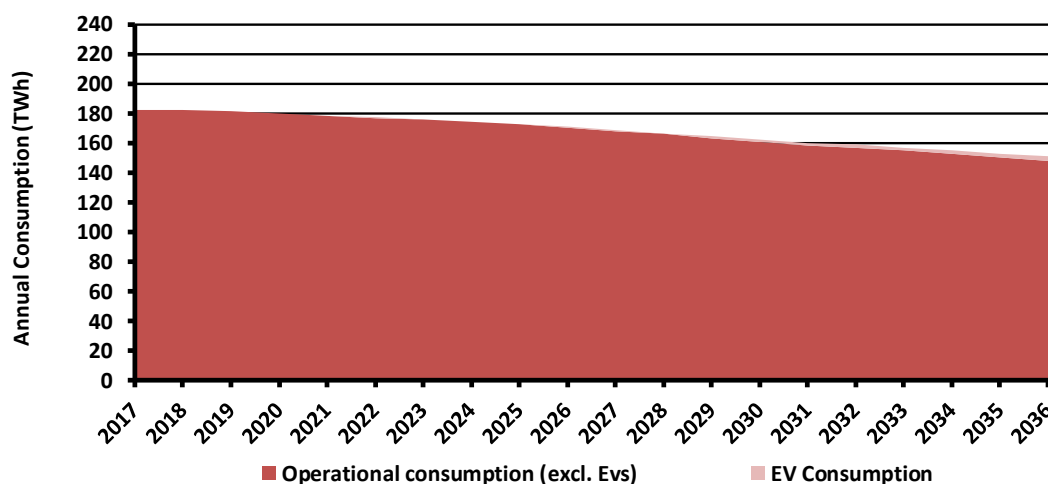


Source:

Energeia

Under the weak sensitivity, EV electricity consumption reaches 3.34 TWh per year and equates to around 2.68% of AEMO’s Insights forecasts for primary load in 2036 as shown in Figure 17 below. This is around 52% below the neutral scenario, consistent with differences in uptake rates between the sensitivities.

Figure 17 – EV Electricity Consumption (NEM Operational, Weak)



Source: Energeia

4.3 EV Maximum Demand Forecasts

Section 4.3 presents Energeia’s forecasts for maximum demand by region from EV charging and assesses the impact of these on the 2017 Insights maximum demand forecasts prepared by AEMO.

Section 4.3.1 describes Energeia’s forecast of aggregate EV demand including controlled EV charging and uncontrolled DC fast charging. The impact on coincident system maximum demand is then described in Section 4.3.2 by adding coincident EV demand to system demand for each half hour.

All of the forecasts present maximum demand in terms of operational requirements (including losses).

4.3.1 EV Maximum Demand

Energeia’s forecast of annual maximum EV demand by region and sensitivity is presented in Table 5 below.

Half-hourly average EV charging profiles by region, charging type and scenario are provided in Appendix B.

The differences in forecast regional results are driven primarily by EV uptake and EV consumption as discussed in Section 4.1 and Section 4.2, respectively. In addition, maximum demand is also influenced by the

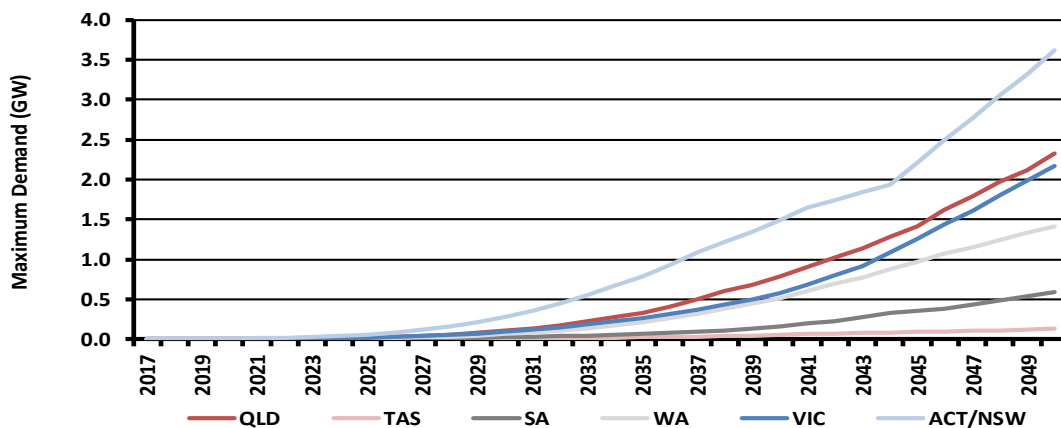
characteristics of each state's peak demand profile, which can impact on the available number of hours available for recharging flexible residential and workplace based EVs. For example, although EV consumption is significantly higher in QLD compared to VIC, there is only a minor difference between EV maximum demand. This is due to differences in peak demand profiles, with QLD's peak demand profile providing a longer period over which to recharge EVs compared to VIC.

Table 5 – EV Maximum Demand (Non-Coincident) by Sensitivity (Operational)

Region	2017			2020			2030			2050		
	EV Max Demand (MW)			EV Max Demand (MW)			EV Max Demand (MW)			EV Max Demand (MW)		
	Strng	Neut	Weak	Strng	Neut	Weak	Strng	Neut	Weak	Strng	Neut	Weak
QLD	0.39	0.38	0.38	5.92	4.44	4.04	301.27	104.90	61.08	4,320.3	2,321.2	943.32
NSW	1.21	1.21	1.20	14.08	10.59	9.95	643.18	278.30	169.4	5,659.2	3,612.7	1,536.0
VIC	0.52	0.52	0.52	4.91	3.96	3.75	224.48	94.52	58.31	3,424.9	2,173.1	619.06
SA	0.12	0.12	0.12	1.45	1.09	1.09	58.20	25.61	16.01	1,119.5	591.81	187.97
TAS	0.03	0.03	0.03	0.50	0.37	0.33	23.86	8.92	5.20	273.57	129.68	59.72
WA	0.19	0.19	0.19	3.51	2.60	2.32	175.67	66.46	38.00	2,658.8	1,415.1	624.97

Source: Energeia

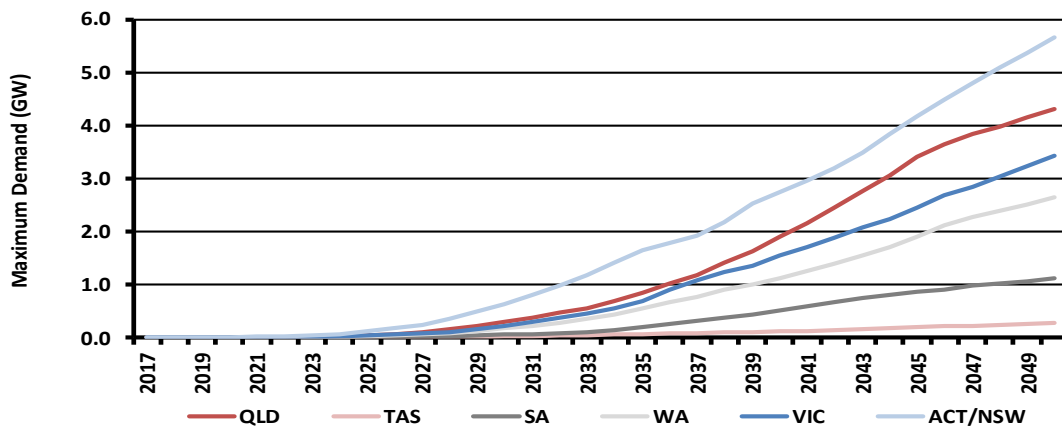
Figure 18 – EV Maximum Demand by Region (Operational, Neutral)



Source: Energeia

Figure 18, Figure 19 and Figure 20 show EV maximum demand by sensitivity. By 2036, forecast EV maximum demand under the strong sensitivity is between 2.9 and 4.2 times greater than under the neutral sensitivity, depending on region, due to the higher EV uptake. By 2050, EV maximum demand reduces to between 160% and 180% larger under the strong sensitivity when compared to the neutral sensitivity.

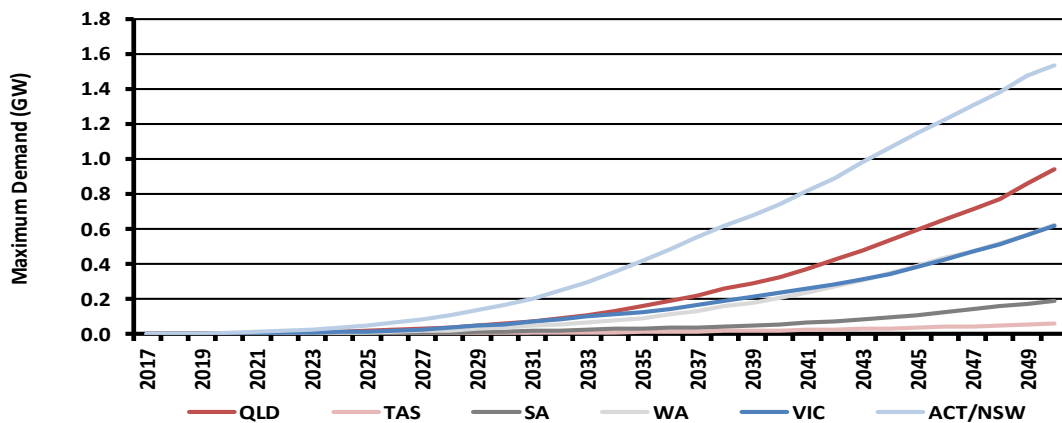
Figure 19 – EV Maximum Demand by Region (Operational, Strong)



Source: Energeia

By 2036, EV maximum demand under the weak sensitivity is between 56% and 65% less than the neutral sensitivity, depending on region, due to lower EV uptake. By 2050, EV maximum demand shifts to between 58% and 62% smaller under the weak sensitivity when compared to neutral.

Figure 20 – EV Maximum Demand by Region (Operational, Weak)

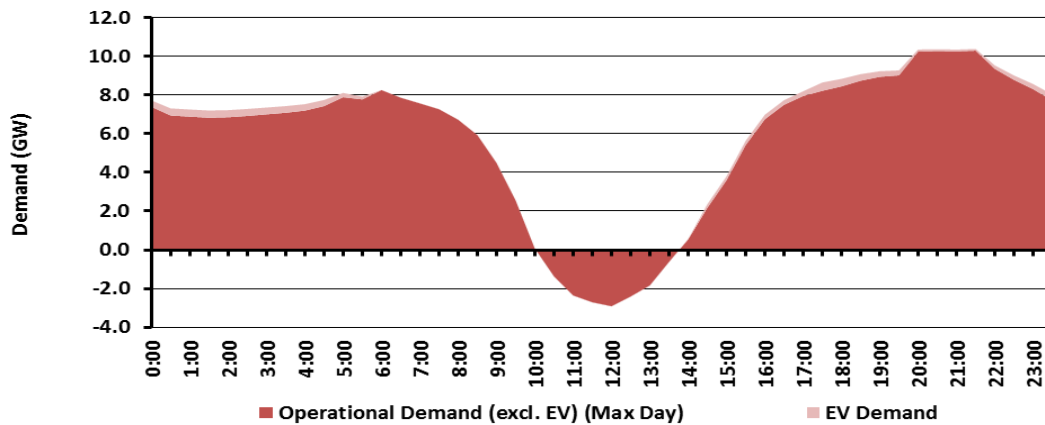


Source: Energeia

4.3.2 Impact on Maximum Demand

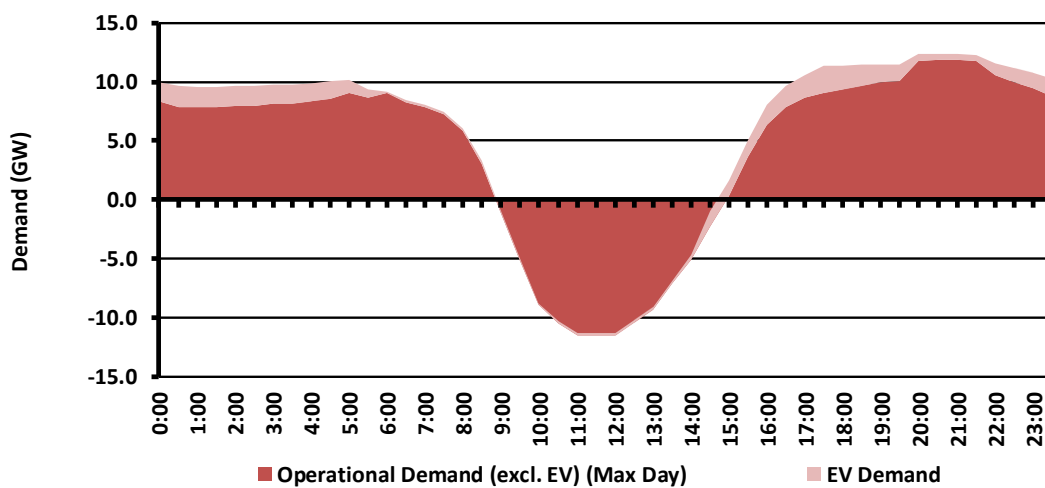
Energeia’s modelling shows EV charging management avoiding increasing maximum demand in any of the regions over the period 2017 to 2036, based on our forecast of underlying operational demand. FiguresFigure 21 to Figure 29 show the contribution of EVs on the maximum demand day for each region for the neutral sensitivity.

Figure 21 – Contribution of EVs on QLD Maximum Demand Day (2036, Neutral)



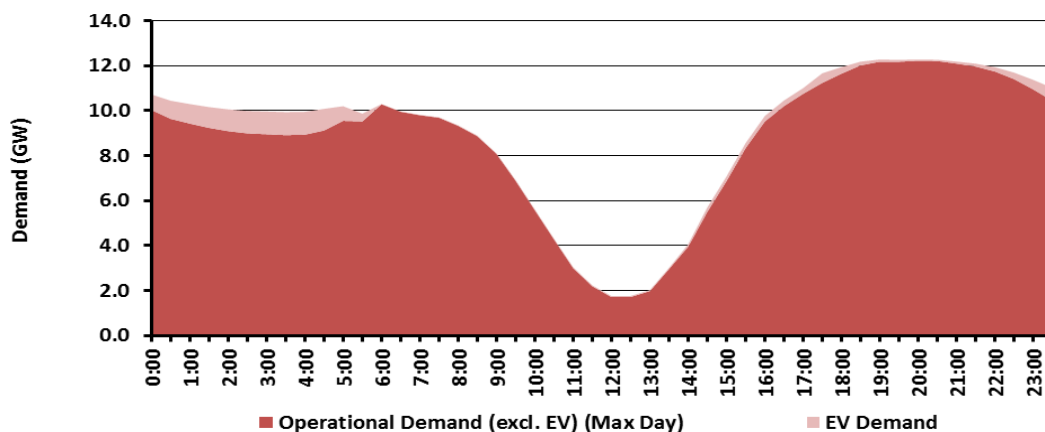
Source: Energeia

Figure 22 – Contribution of EVs on QLD Maximum Demand Day (2050, Neutral)



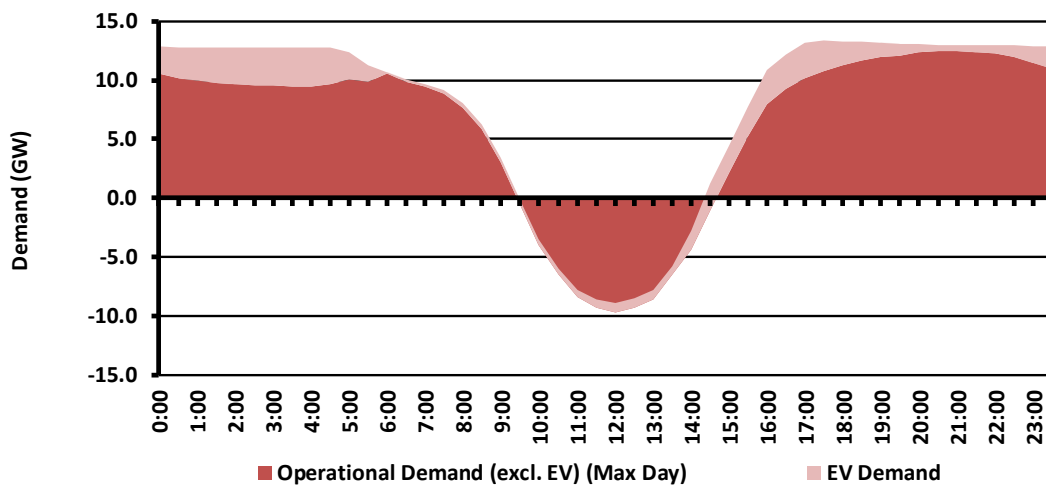
Source: Energeia

Figure 23 – Contribution of EVs on NSW Maximum Demand Day (2036, Neutral)



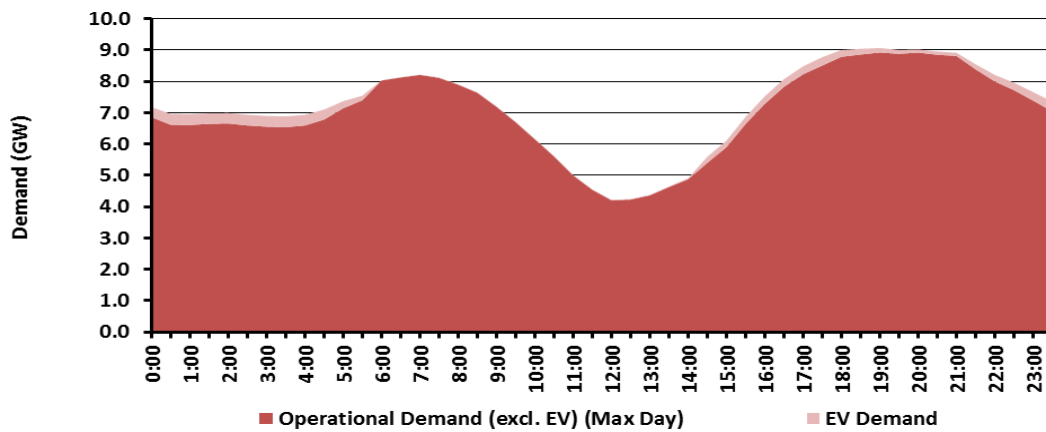
Source: Energeia

Figure 24 – Contribution of EVs on NSW Maximum Demand Day (2050, Neutral)



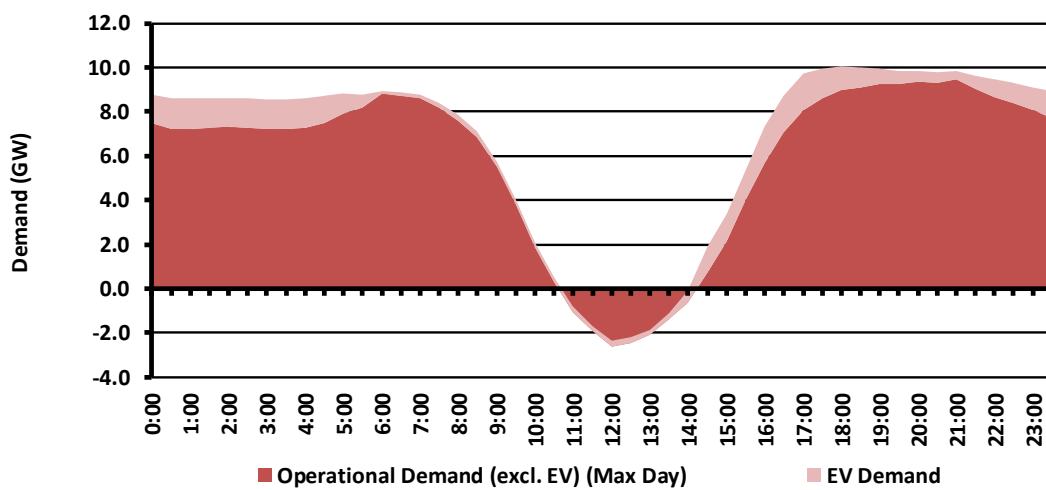
Source: Energeia

Figure 25 – Contribution of EVs on VIC Maximum Demand Day (2036, Neutral)



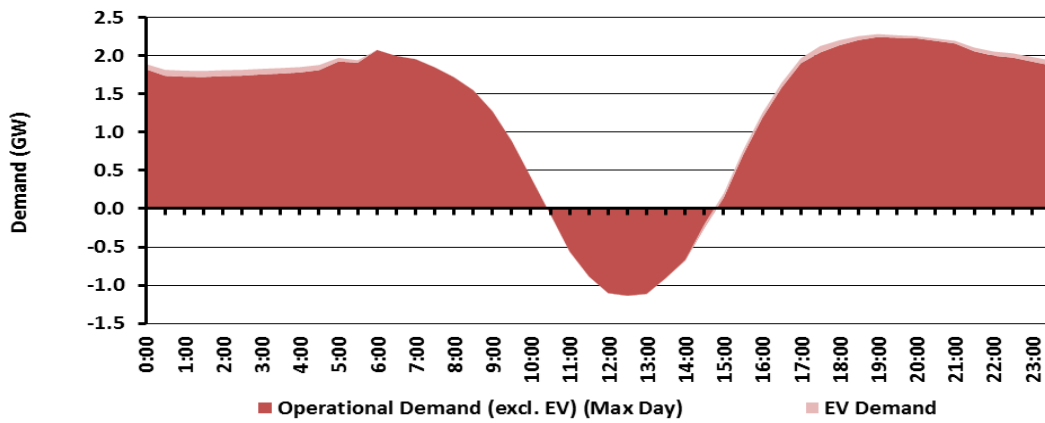
Source: Energeia

Figure 26 – Contribution of EVs on VIC Maximum Demand Day (2050, Neutral)



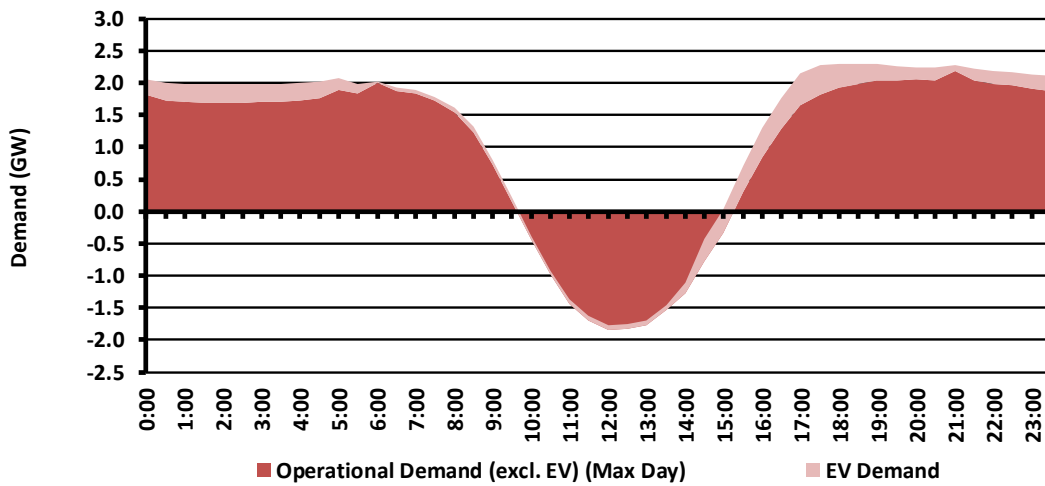
Source: Energeia

Figure 27 – Contribution of EVs on SA Maximum Demand Day (2036, Neutral)



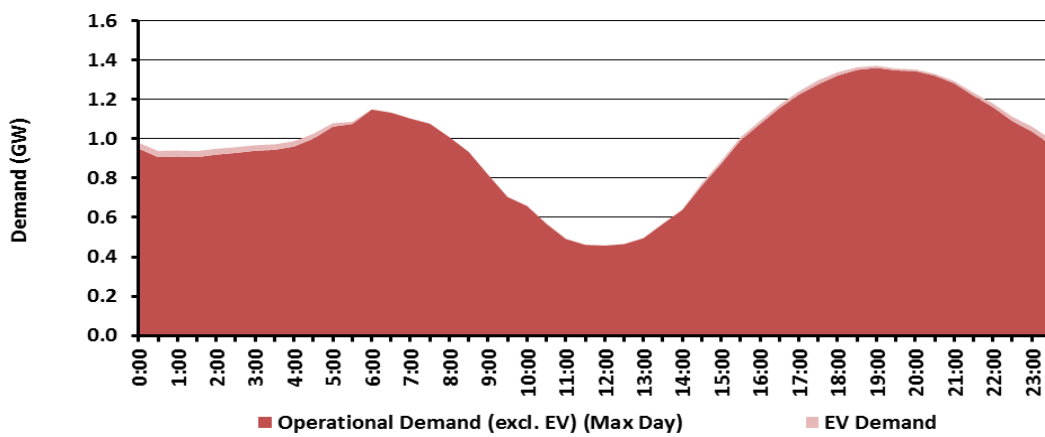
Source: Energeia

Figure 28 – Contribution of EVs on SA Maximum Demand Day (2050, Neutral)



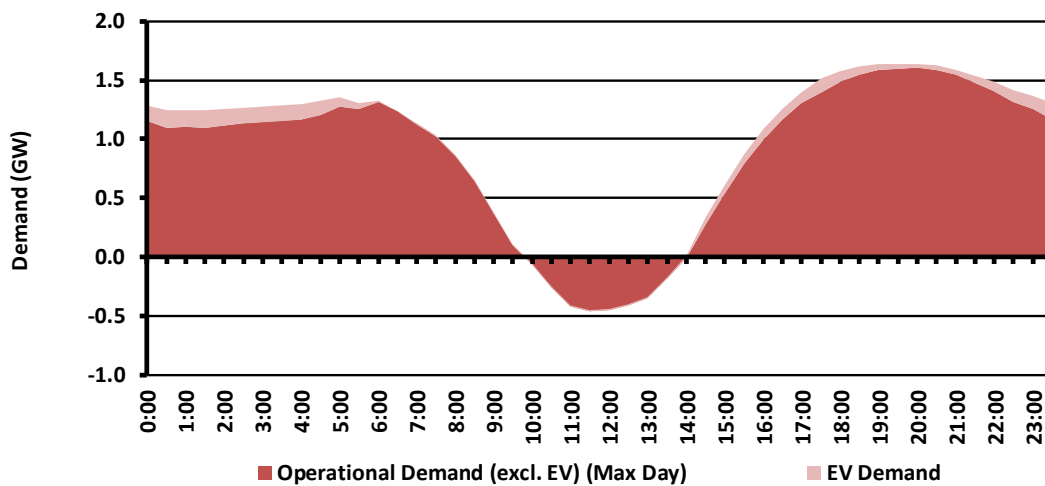
Source: Energeia

Figure 29 – Contribution of EVs on TAS Maximum Demand Day (2036, Neutral)



Source: Energeia

Figure 30 – Contribution of EVs on TAS Maximum Demand Day (2050, Neutral)

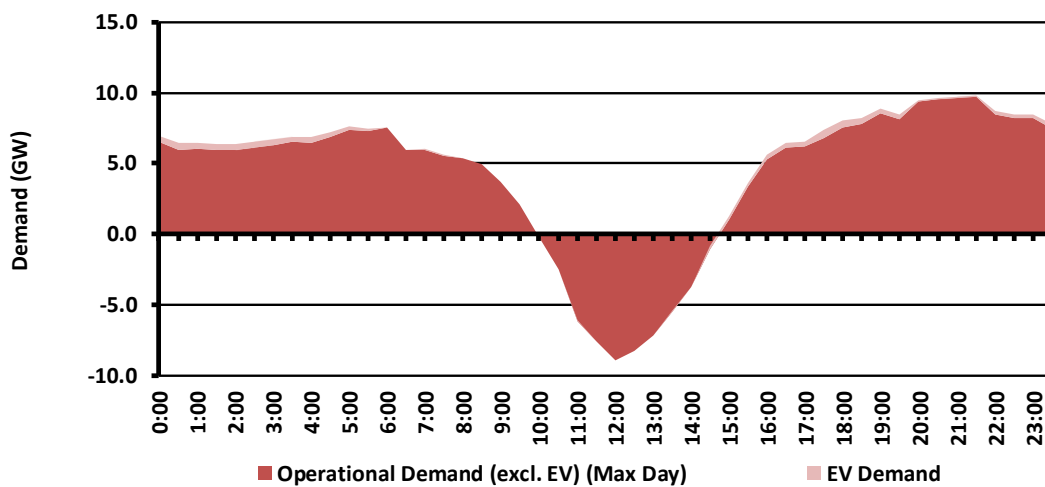


Source: Energeia

4.3.3 Impact on Minimum Demand

Energeia’s modelling of future minimum demand shows a shift from overnight to the middle of the day by 2036 in all states as solar PV penetration increases. As a result, daytime EV charging is forecast to increase the minimum demand for all of the regions by 2036. DCFC of vehicles slightly increases minimum demand, although the effect is almost insignificant due to the majority of charging still occurring overnight as shown in Figures Figure 31 to Figure 39 Fig (although the effect on minimum demand is too small to be seen).

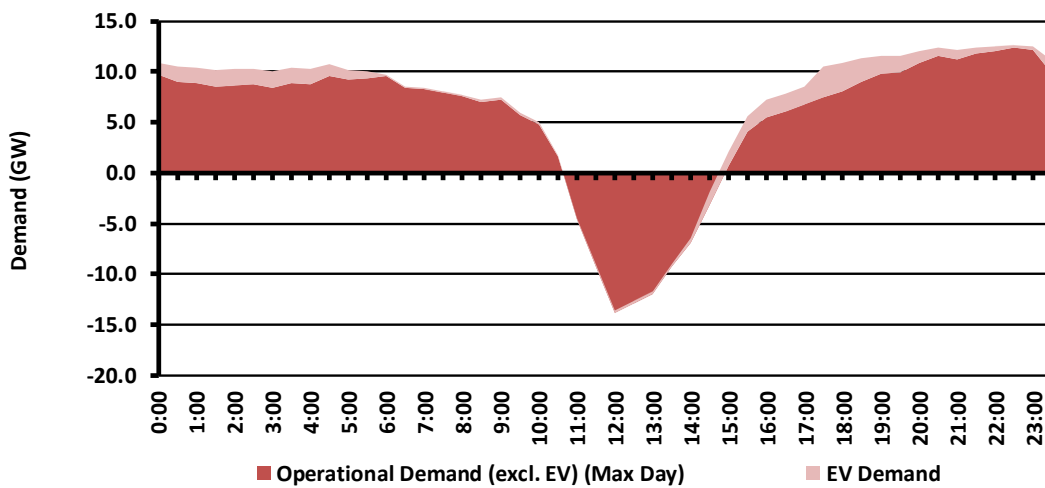
Figure 31 – Contribution of EVs on QLD Minimum Demand Day (2036, Neutral)



Energeia

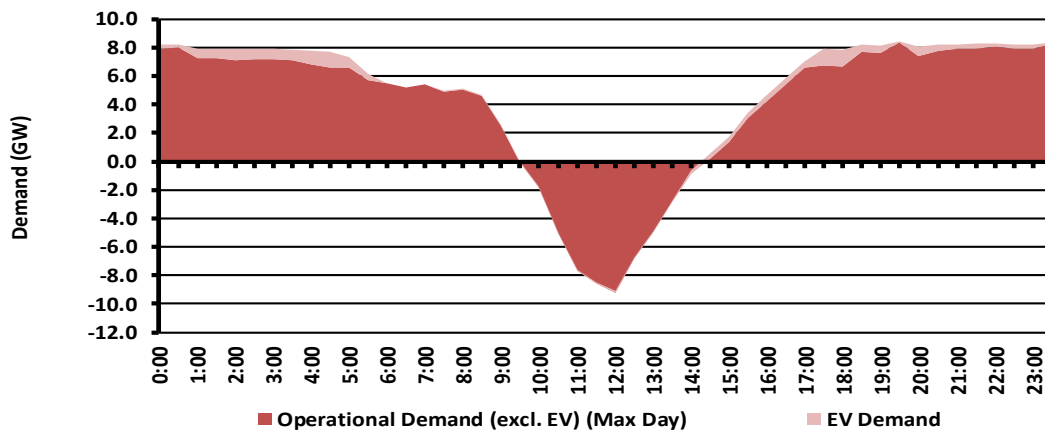
Source:

Figure 32 - Contribution of EVs on QLD Minimum Demand Day (2050, Neutral)



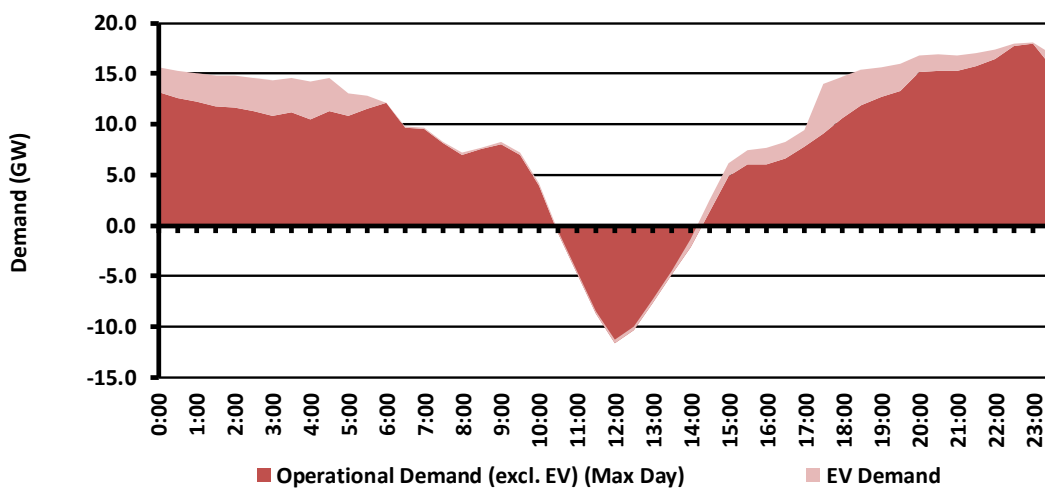
Source: Energeia

Figure 33 – Contribution of EVs on NSW Minimum Demand Day (2036, Neutral)



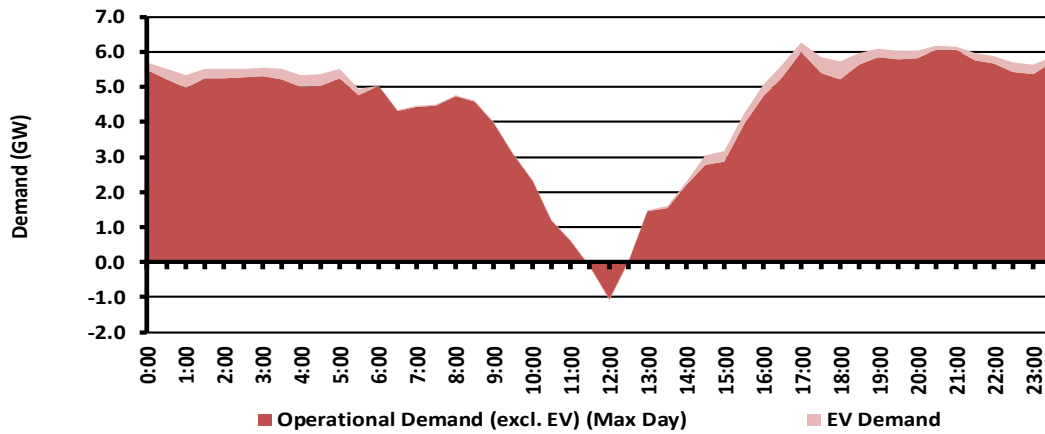
Source: Energeia

Figure 34 – Contribution of EVs on NSW Minimum Demand Day (2050, Neutral)



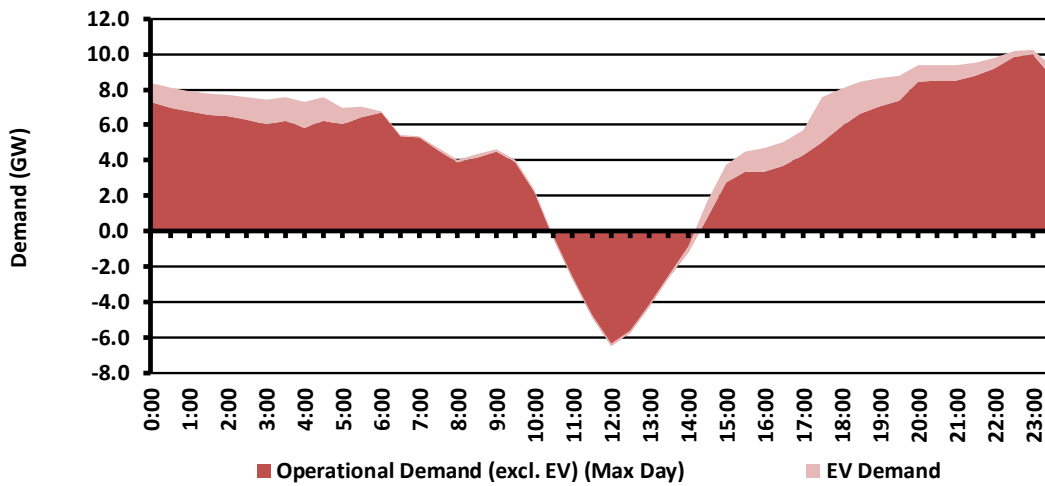
Source: Energeia

Figure 35 – Contribution of EVs on VIC Minimum Demand Day (2036, Neutral)



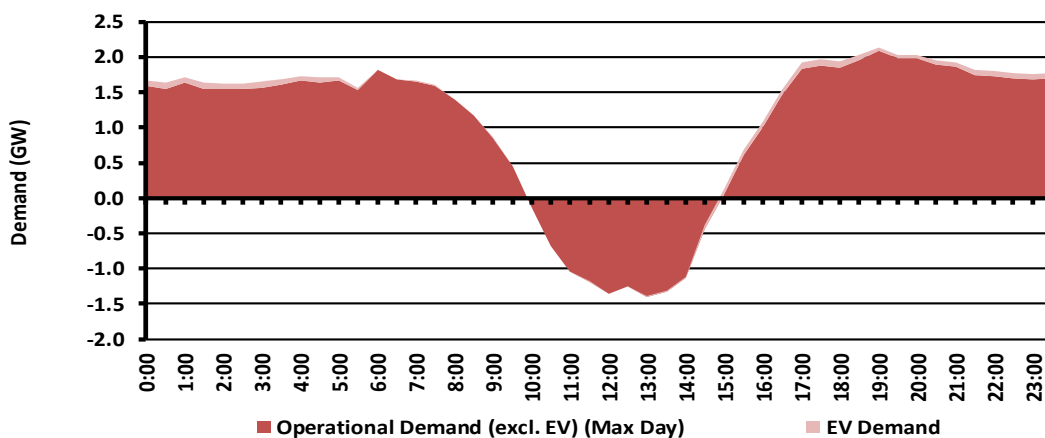
Source: Energeia

Figure 36 – Contribution of EVs on VIC Minimum Demand Day (2050, Neutral)



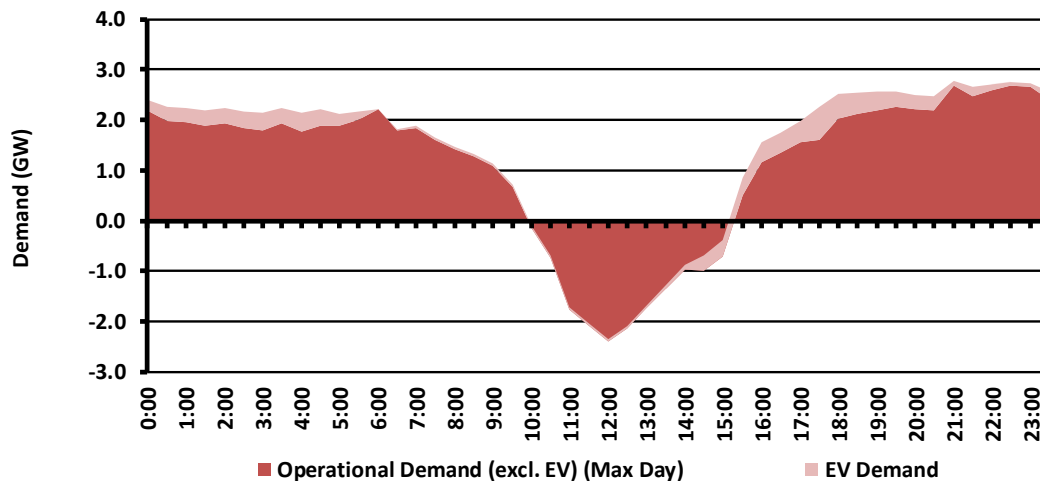
Source: Energeia

Figure 37 – Contribution of EVs on SA Minimum Demand Day (2036, Neutral)



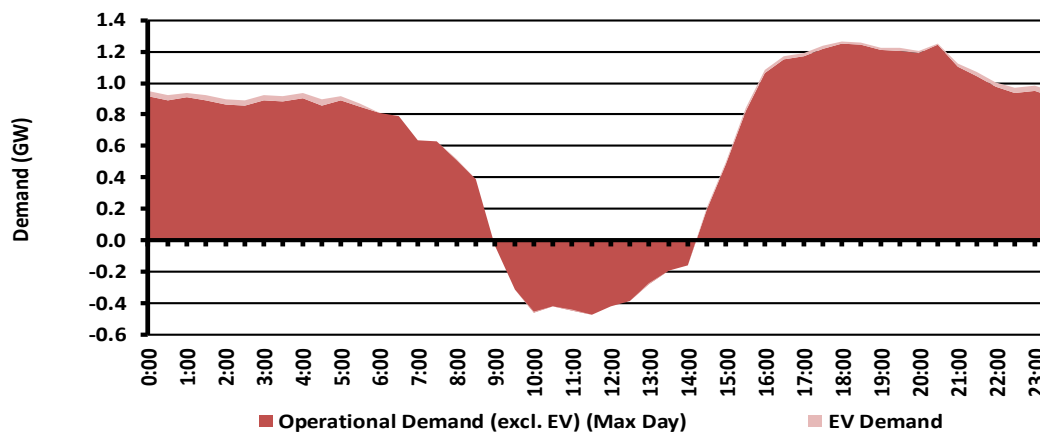
Source: Energeia

Figure 38 – Contribution of EVs on SA Minimum Demand Day (2050, Neutral)



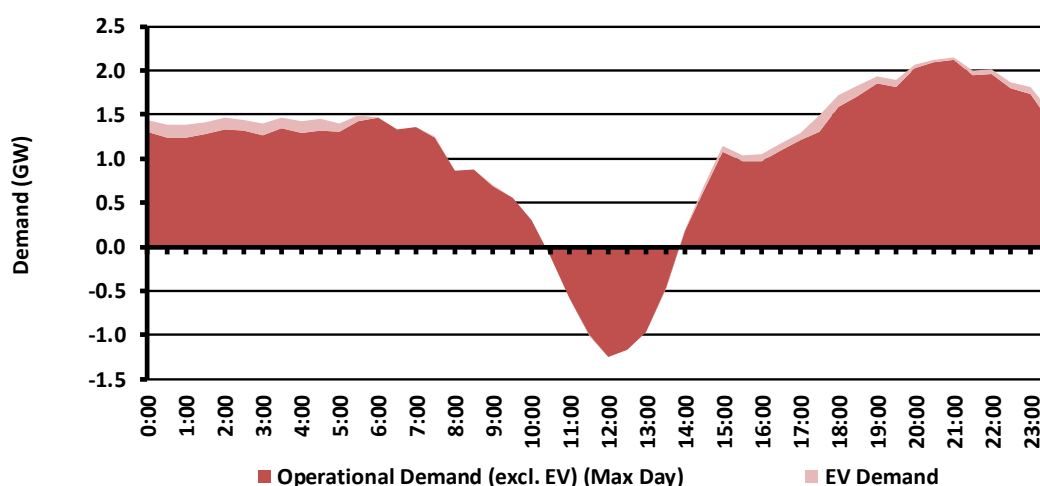
Source: Energeia

Figure 39 – Contribution of EVs on TAS Minimum Demand Day (2036, Neutral)



Source: Energeia

Figure 40 – Contribution of EVs on TAS Minimum Demand Day (2050, Neutral)



Source: Energeia

While the above results suggest an opportunity to encourage daytime EV charging, a whole of system analysis is required to determine the optimal EV charging profile. A key issue to be addressed is whether higher charging in the middle of the day will increase peak demand and augmentation costs for distribution networks, especially in commercial areas where peak demand is already set in the middle of the day.

5 Recommendations for Future Modelling

The EV forecasts contained throughout this paper have been prepared to provide insight into the potential impact of future EV uptake on AEMO forecasts. Energeia's EV forecasts have been prepared based on existing publicly available data. In future years, it is understood that AEMO intends to integrate EV forecasting into the Electricity Forecast Insights process to allow for greater feedback between the primary forecast and EV uptake drivers.

Accordingly, there are a range of key uncertainties and limitations which Energeia recommends to address and/or improve in future modelling.

5.1 Key Uncertainties

The EV forecasts within this paper contain a number of key uncertainties which affect the precision and accuracy of the results. These include:

- The structure of tariffs to be applied to EV charging and changes in these structures over time (See Section 5.2 below)
- Policy uncertainty, with respect to:
 - The mechanism and timing of introduction of a vehicle greenhouse gas emission standard
 - The application of a broader carbon price to vehicle emissions.
- The rate at which vehicle manufacturers make EV models available within the Australian market (nominated as model availability within this paper)
- The location, number and duty cycle of fast charging
- Original Equipment Manufacturer (OEM) response to Vehicle Emission Standards (VES)
- Government policy or subsidy to incentive EV uptake.

Further, the near-term EV forecasts are subject to a high degree of uncertainty due to the immaturity of the market and the short-term actions that may be taken by the private sector to accelerate uptake. For example, there is the potential for early action by industry to promote EVs via heavily subsidised tariffs⁷.

In addition, there are likely to be further drivers, external to the model, relating to substitutable low emission technologies by OEMs, including natural gas and fuel cell vehicles. Consideration of the potential impacts of these have not been considered within the model in terms of the extent to which new technologies are likely to limit EV model availability. That is, the model assumes that a wholesale transition of the Australian vehicle fleet to EVs will occur due to lower operating costs, better vehicle environmental and acceleration performance and higher safety ratings for EVs when compared to their counterparts.

The impact of self-driving cars or wireless induction charging have not been considered by the model.

5.2 Changes in EV Charging Tariffs over Time

The EV forecasting model assumes that charging management change over time to compensate for shifting peak demand times from DER and increased penetration of EVs. All home charging and commercial vehicle charging is assumed to be managed by 2036.

Whether or not managed EV charging is operating during the middle of the day to mitigate excess generation of solar PV will depend on the net benefits across the industry, including the distribution, transmission and generation sectors. Whole-of-system cost-to-serve analysis is needed to determine what the optimal shape will ultimately be over time.

⁷ See for example, Vesey, Andy (@AndyVesey_AGL) "\$1 a day (fully carbon offset) to charge your #EV. AGL to launch Nov 2016. \$365 pa max for all your EV trips #AEW16" 5:37 PM, 20 June 2016. Tweet.

It is recommended that for future EV modelling, AEMO considers modelling a whole of system cost-to-serve to enable the least cost demand profile for EVs to be identified and incorporated into future EV load forecasts.

5.3 Integration with Primary Load

The forecasts assume that the decision to adopt an EV is made independently from any other decisions regarding primary energy consumption. In reality, there will be a subset of customers for whom the decision to purchase an EV could be made more attractive if combined with a solar PV system depending on the tariff arrangements and individual driving patterns.

Further, the present modelling assumes that the EV is not capable of any vehicle to home or vehicle to grid (V2G) charging. Where this is the case, integration with the primary load becomes critical to residential forecasts and interacts with the stationary storage uptake.

While AEMO has separately undertaken solar PV and battery storage forecasts, these have not been integrated with the EV forecasts in this paper.

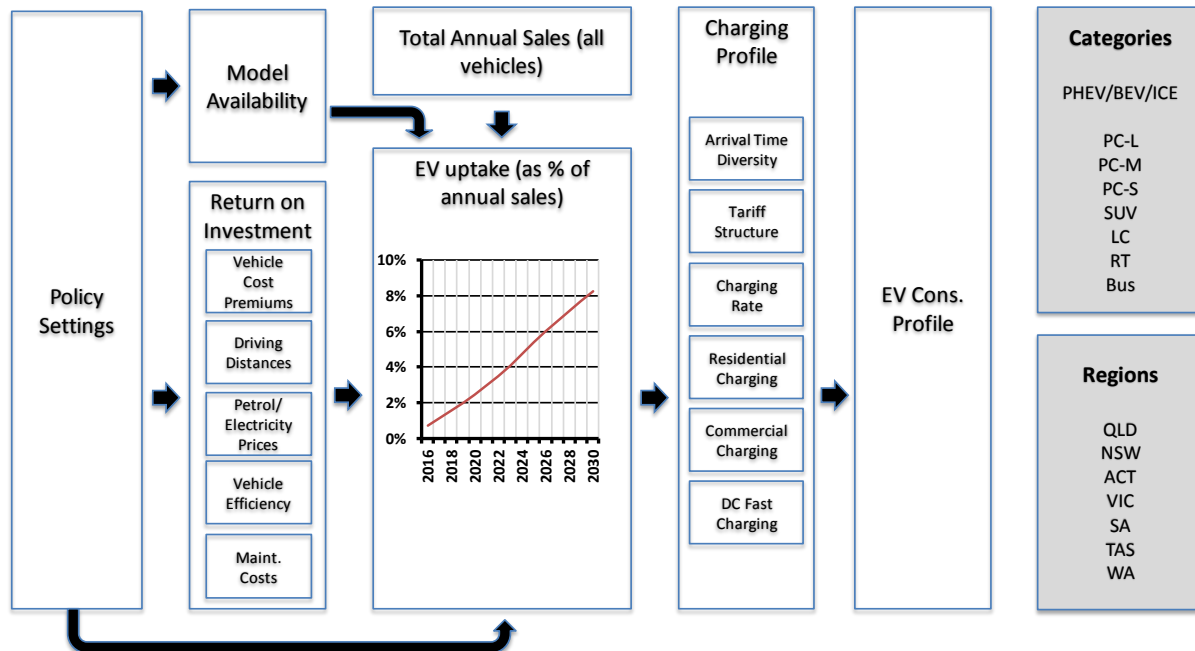
It is recommended that for future EV modelling, AEMO integrates the EV uptake and charging decisions with the broader customer decision making with respect to solar PV and stationary battery storage uptake and operation.

Appendix A: Detailed Assumptions

A.1 Overview of Model Approach

Energeia's EV forecasting model is comprised of two parts, EV uptake and EV charging as shown in Figure 41 below.

Figure 41 – Energeia EV Forecasting Model



Source: Energeia

The EV uptake module forecasts EV uptake as a percentage of annual vehicle sales for each category of vehicle type by state. This is based on vehicle model availability and the vehicle owner's return on investment.

The EV charging module then applies a charging regime to each vehicle adopted based on the arrival and departure time of the vehicle at the point of charge, the number of kilometres travelled and the least cost demand profile (managed charging only).

A.2 EV Uptake

EV uptake is determined by a two-parameter function that describes vehicle uptake over time based on:

1. EV premium payback more than two years:

$$EV\ Uptake_t = Total\ New\ Vehicle\ Sales_t * (a_t \times ROI_t + b_t \times Model\ Availability_t)$$

2. EV premium payback less than two years (tipping point):

$$EV\ Uptake_t = Total\ New\ Vehicle\ Sales_t * MIN(Upper\ EV\ Limit, Model\ Availability_t)$$

Where:

- $Total\ New\ Vehicle\ Sales_t$ = Total new vehicle sales within a given vehicle class in year t
- $Model\ Availability_t$ = Percentage of models within a given vehicle class available in EV form in year t. This inclusion of this factor reflects that, for the mass market, a primary driver of vehicle purchase is the availability of that model in EV form. This factor effectively places an upper bound on EV adoption, which is determined by a scenario based parameter.

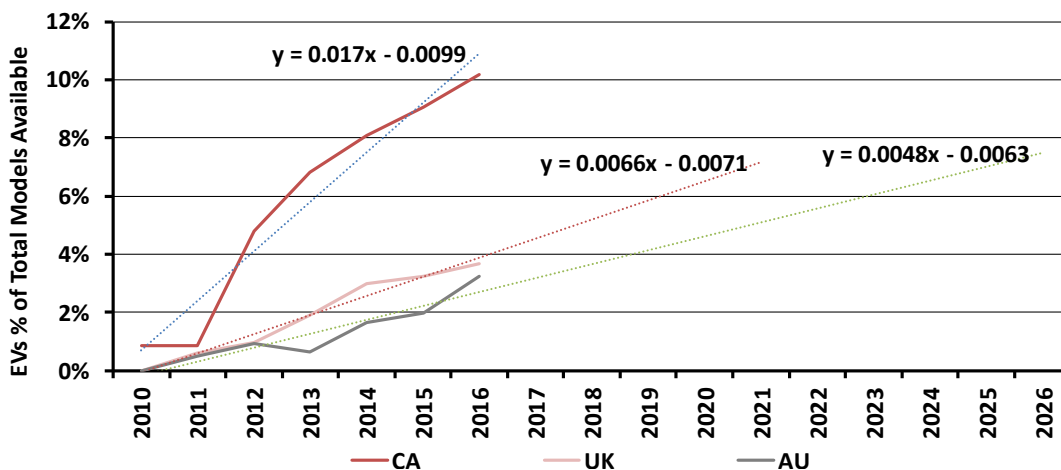
- *Upper EV Limit* = Upper model availability limit for all vehicles within a given vehicles class
- ROI_t = The first-year return on investment for the vehicle owner investing in an EV in year t in terms of reduced operational costs (fuel) and premium paid compared to the equivalent ICE vehicle
- a_t = Model coefficient derived from historical data of diesel and hybrid electric vehicle uptake for observed ROIs
- b_t = Model coefficient derived from historical data of diesel and hybrid electric vehicle uptake for observed model availability

As seen, EV uptake depends on the functional form assumed for model availability and change in ROI over time. It should be noted that Energeia’s ROI calculation does not take into account step changes in depreciation or salvage value due to increasing EV penetration. These factors are explained in further detail below.

A.2.1 Incentives Impact on Model Availability

Energeia has developed its assumed rate of EV model availability based on an empirical analysis of model availability relative to the level of jurisdictional incentives. Figure 42 below displays the results of our analysis of the UK, California and Australian markets. It shows that California, the market with the highest EV incentive at around \$10,000 USD including Federal incentives, sees the fastest rate of new EV model introductions. The UK market, which offers around \$5,000 USD in incentives, is higher than virtually incentive-free Australia.

Figure 42 – EV Model Availability by Year by Key Market



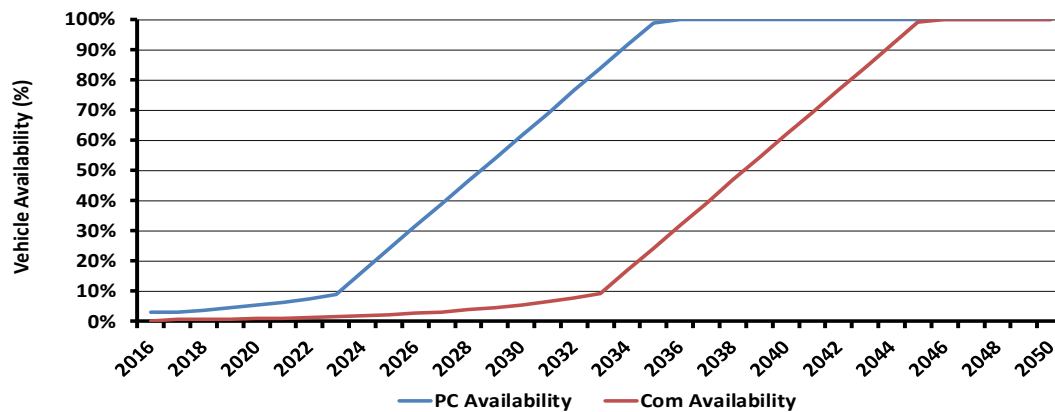
Source: Energeia

The above analysis was used to develop an EV model introduction function based on the level of assumed incentive. Scenarios with incentives comparable to California see OEM introducing new EV models at the California rate, while scenarios with incentives closer to zero see new EV models introduced at the historical Australian rate, as shown in the figure above.

A.2.2 Assumed Model Availability

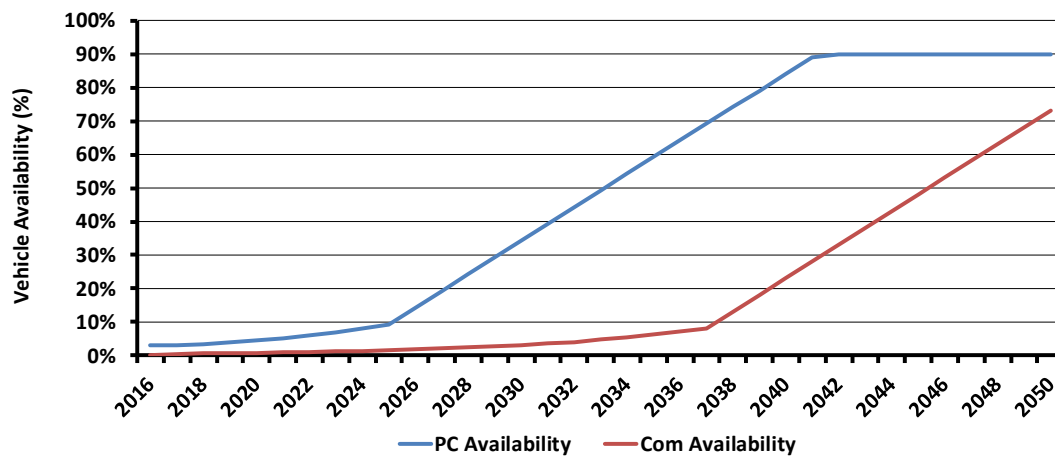
Assumed model availability varies by vehicle class and by sensitivity. A 9% of model threshold trigger for EVs hitting a maximum ramp rate shown in Figure 43, Figure 44 and Figure 45 is based on research and analysis of international EV model availability ramp rates given varying incentives over time and by region.

Figure 43 – Model Availability Strong Sensitivity



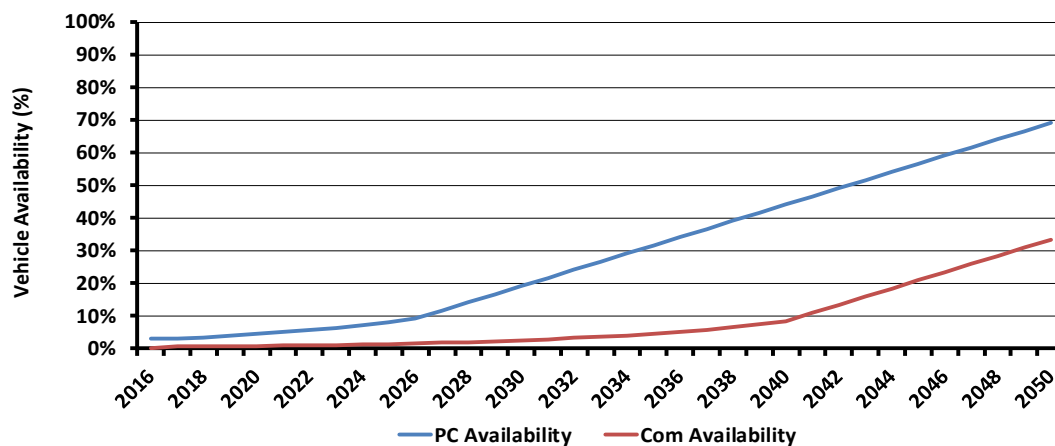
Source: Energeia

Figure 44 – Model Availability Neutral Sensitivity



Source: Energeia

Figure 45 – Model Availability Weak Sensitivity



Source: Energeia

A.2.4 Impacts of Obsolescence on Vehicle Depreciation

Energeia analysed two datasets in order to estimate the link between new vehicle technology uptake and accelerated depreciation rates of obsolete technology. The analysis looked at data on the resale value of vehicle

models where new technology penetration (diesel) was clearly becoming dominant, as well as data on the resale value of vehicle models following the OEM's discontinuation of the model.

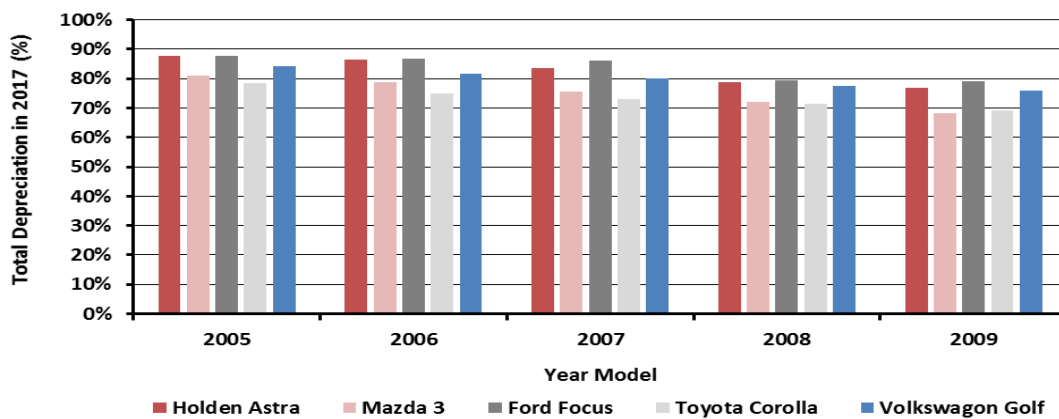
Based on this analysis, Energeia has excluded depreciation as a material input into our calculation of EV ROIs.

OEM Discontinuation Analysis

The first study compared the cumulative depreciation of discontinued models against themselves (before they were discontinued) and against comparable models.

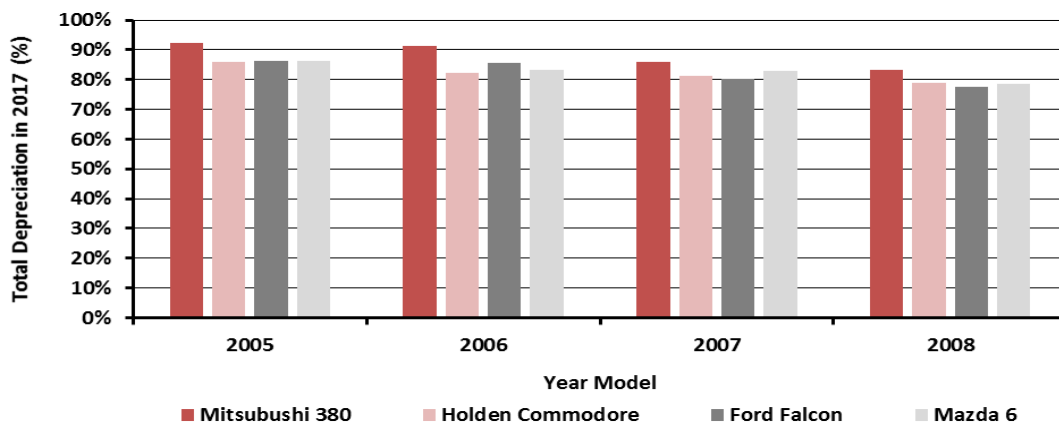
The analysis shows that even though the Astra was discontinued in 2009, the level of depreciation of the 2009 model shows only a slightly faster rate of depreciation when compared to similar sized, non-discontinued models. Analyses of other examples confirms the result. Discontinuation does not lead to significantly different rates of depreciation, with the level of depreciation between the runout and continuous remaining similar, within +/-10%.

Figure 46 – Holden Astra (2009) Runout



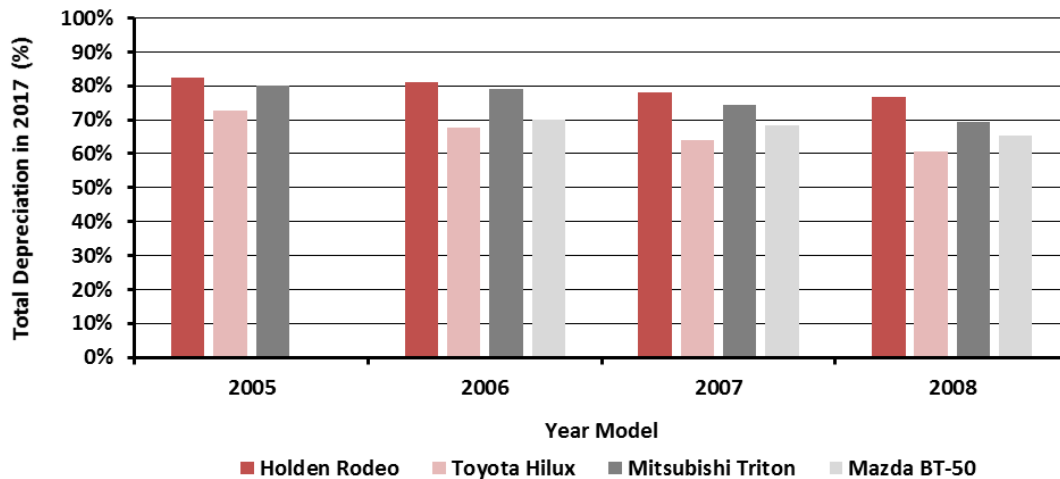
Source: VFACTS, Redbook

Figure 47 – Mitsubishi 380 (2008) Runout



Source: VFACTS, Redbook

Figure 48 – Holden Rodeo (2008) Runout

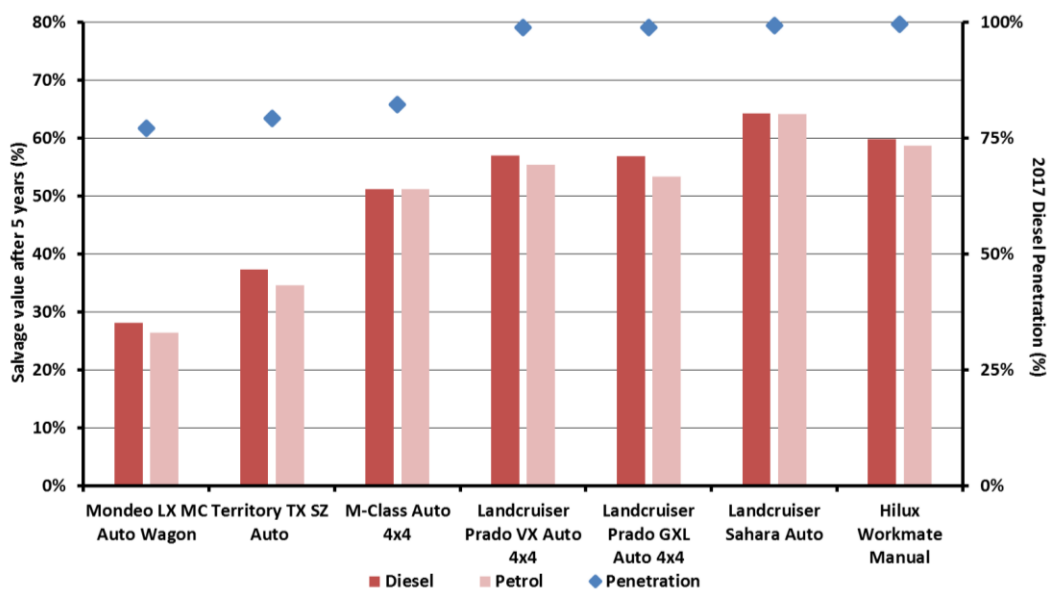


Source: VFACTS, Redbook

Market Obsolescence Analysis

The second analysis compared the resale value of ICE and diesel versions of the same vehicle model after 5 years in cases where diesel ultimately reached over 75% of the market. The results show a +/- 5% difference between the unpopular petrol and popular diesel models. Energeia speculates that this is because there remains a small but committed market niche that continues to favour the petrol model for whatever reason.

Figure 49 – Salvage value of petrol and diesel models after 5 years



Source: VFACTS, Redbook

Based on the above analysis, Energeia has concluded that model obsolescence does not significantly impact on vehicle resale value, all else being equal.

A.3 Operation and Maintenance Costs

A.3.1 Electricity Tariffs

The model assumes the EVs are charged on a controlled load tariff or with DCFCs. The tariffs described in Table 6 are used in the model and are not sensitivity dependent.

Table 6 – Electricity Tariffs

State	2017 Retail Price (\$/kWh)
QLD	\$0.1368
NSW	\$0.1241
VIC	\$0.1669
SA	\$0.2063
TAS	\$0.1160
WA	\$0.1132

A.3.2 Electricity Price

Both the retail and network components of EV charging tariffs are grown over time in the EV uptake model and vary by state and by sensitivity. The model uses the retail electricity price projections developed by Jacobs for AEMO⁸ in real terms.

The electricity price trend has a direct impact on EV fuel expenditure.

⁸ Jacobs, Retail electricity price history and projections – Public, June 2017

A.3.3 Petrol Price

Energeia’s petrol price forecasts have been developed using historical relationships between the price of petrol and the oil price, which are then projected using the sensitivity assumption for oil prices. Our assumed inputs by state, year and sensitivity are shown in Table 7 below.

Table 7 – Fuel Price by State

Year	Low						Neutral						High					
	WA	QLD	SA	TAS	ACT/NSW	VIC	WA	QLD	SA	TAS	ACT/NSW	VIC	WA	QLD	SA	TAS	ACT/NSW	VIC
2017	\$1.15	\$1.15	\$1.14	\$1.21	\$1.15	\$1.14	\$1.15	\$1.15	\$1.14	\$1.21	\$1.15	\$1.14	\$1.15	\$1.15	\$1.14	\$1.21	\$1.15	\$1.14
2018	\$1.14	\$1.13	\$1.12	\$1.19	\$1.13	\$1.12	\$1.17	\$1.17	\$1.16	\$1.22	\$1.17	\$1.16	\$1.21	\$1.20	\$1.19	\$1.26	\$1.20	\$1.19
2019	\$1.12	\$1.11	\$1.10	\$1.17	\$1.11	\$1.10	\$1.19	\$1.18	\$1.17	\$1.24	\$1.18	\$1.17	\$1.26	\$1.25	\$1.24	\$1.31	\$1.25	\$1.24
2020	\$1.10	\$1.09	\$1.08	\$1.15	\$1.09	\$1.08	\$1.20	\$1.20	\$1.19	\$1.26	\$1.20	\$1.19	\$1.31	\$1.30	\$1.29	\$1.37	\$1.31	\$1.29
2021	\$1.08	\$1.07	\$1.06	\$1.13	\$1.08	\$1.06	\$1.22	\$1.21	\$1.20	\$1.28	\$1.22	\$1.20	\$1.36	\$1.35	\$1.34	\$1.42	\$1.36	\$1.34
2022	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2023	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2024	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2025	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2026	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2027	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2028	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2029	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2030	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2031	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2032	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2033	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2034	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2035	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2036	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2037	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2038	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2039	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2040	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2041	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2042	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2043	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2044	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2045	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2046	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2047	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2048	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2049	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39
2050	\$1.06	\$1.05	\$1.05	\$1.11	\$1.06	\$1.05	\$1.24	\$1.23	\$1.22	\$1.29	\$1.23	\$1.22	\$1.41	\$1.41	\$1.39	\$1.48	\$1.41	\$1.39

Source: Energeia

A.3.4 Travel Distance

The travel distance dictates energy requirements and therefore has a direct impact on both ICE vehicles and EV annual fuel expenditure. The model adopts an average driving distance in this application to determine annual vehicle costs that vary by state and by vehicle class as summarised in Table 8.

Table 8 – Travel Distance

State	Annual Average Distance Travelled (km/year)	
	Light Passenger	Light Commercial
NSW	12,300	17,100
ACT	12,800	18,200
VIC	13,800	17,700
QLD	13,300	17,100
SA	11,600	16,700
WA	12,400	17,200
TAS	11,600	12,100

Source: ABS Survey of Motor Vehicle Use

A.3.5 Fuel Consumption

Fuel efficiency in the model is a key factor in determining energy requirements and fuel costs. The underlying fuel efficiency of ICE vehicles and EVs stay constant in the model as combustion and electric engines are well understood and established technologies.

The assumptions for fuel consumption are summarised in Table 9. These estimates have been developed based on OEM reported efficiency data.

Table 9 - Fuel Consumption

Vehicle Type	2017 Efficiency	
	EV kWh/km	ICE L/km
Passenger Car Small	0.137	0.052
Passenger Car Medium	0.178	0.063
Passenger Car Large	0.181	0.102
Sport Utility Vehicle Medium	0.181	0.064
Sport Utility Vehicle Large	0.181	0.104
Light Commercial	0.155	0.065
Rigid Truck	0.400	0.488
Bus	0.364	0.445

Source: Energeia, OEM websites

A.4 Capital Costs

The vehicle purchase price is broken down into three components in the model as shown in Table 10.

Table 10 – Capital Cost

Cost Component	ICE	BEV	PHEV
Balance of System	✓	✓	✓
Battery		✓	✓

The balance of system of a vehicle encompasses all the components of the vehicle other than the EV batteries. Each of the above components is described in the following sections.

A.4.1 EV Premium

The model assumes an EV premium costs described in Table 11 in 2017. These estimates have been developed based on OEM reported efficiency data.

Table 11 – EV Premium

Vehicle Class	EV Premium
Passenger Car Small	\$ 30,110
Passenger Car Medium	\$ 15,500
Passenger Car Large	\$ 22,805
Sport Utility Vehicle Medium	\$ 2,398
Sport Utility Vehicle Large	\$ 5,689
Light Commercial	\$ 11,010
Rigid Truck	\$ 42, 229
Bus	\$ 583, 463

Source: Energeia and OEM websites

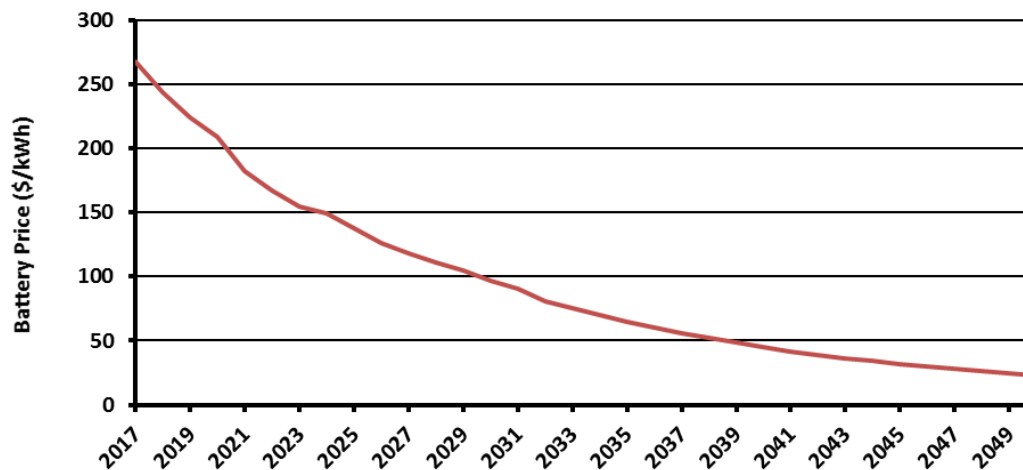
Premiums reduce over time by scenario as detailed in in Section 3.2.

A.4.2 Battery Cost

Energeia's short and medium term battery price outlook is a function of expected improvements in lithium based manufacturing and economies of scale, while the long-term battery price outlook is based on next generation storage technologies that will achieve higher energy densities with significantly less raw material.

The model assumes a decline in lithium price over the modelling period leading to the battery cost projection shown in Figure 50. This forecast is based on a consensus average among leading international lithium price forecasters. The battery price does not vary with sensitivity.

Figure 50 – EV Storage Price Outlook



Source: Energeia research

A.5 EV Charging

The EV charging profile is determined by aggregating the unique charging profile of each individual electric vehicle adopted. The individual profiles are assigned based on:

- Whether the vehicle is assigned as L2 (9.6kW) home charging, L2 commercial charging (charges at work or depot location), or Direct Current Fast Charging (DCFC) which is defined as the EV equivalent of a gas station (1MW station with 5 min charge time)
- DCFC chargers enable drivers without a garage to own an EV, encourage EV charging during hours of excess supply from solar PV, extend EV range to enable EV use for any trip type
- The daily travel distance for both weekday and weekend travel (drawn from a database of regionally specific diversified travel distances), which determines the amount of charge to be supplied by day type
- An arrival time for both weekday and weekend travel (drawn from a database of diversified times specific to either home charging or commercial charging) which dictates when charging starts, in the absence of any other tariff restrictions
- A departure time for both weekday and weekend travel (drawn from a database of diversified times specific to either home charging or commercial charging) which dictates when charging must cease in the absence of any other tariff restrictions
- For home and workplace charging, the optimal EV weekday and weekend demand profile for a given state to minimise whole-of-system cost.
- For DCFC charging, the weekday and weekend DCFC demand profile is based on the weekday and weekend transportation demand profile.

A.5.1 Type of Charging

A vehicle can be assigned to either a L2 home charger, a L2 commercial charger or DCFC.

Passenger cars allocated to DCFC reflect the percentage of households in each state with more than one vehicle. Energeia expects these vehicles will use DCFC rather than try and share private parking space. Commercial vehicles are assumed to be charged at their respective depots.

Detailed charge type assumptions are shown in Table 12.

Table 12 - Charger Type

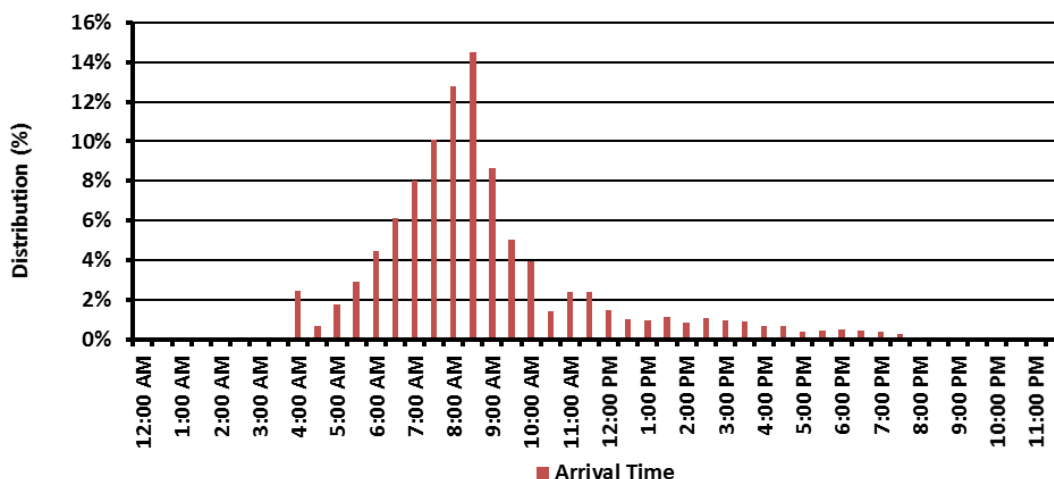
Vehicle Type	Charger Type	NSW	QLD	SA	VIC	WA	TAS
Residential	Destination (Home) Charging	41.2%	37.8%	35.3%	27.5%	15.5%	37.2%
	DCFC Public Charging	58.8%	62.2%	64.7%	72.5%	84.5%	62.8%
Commercial	Destination (Home and Depot) Charging	100%	100%	100%	100%	100%	100%

Source: Energeia

A.5.3 Destination Charging Start Times

The charging start time constraint for each managed charging EV is determined by the vehicle arrival time. The model uses the arrival time distribution shown in Figure 51.

Figure 51 – Vehicle Arrival Distribution by Vehicle Type



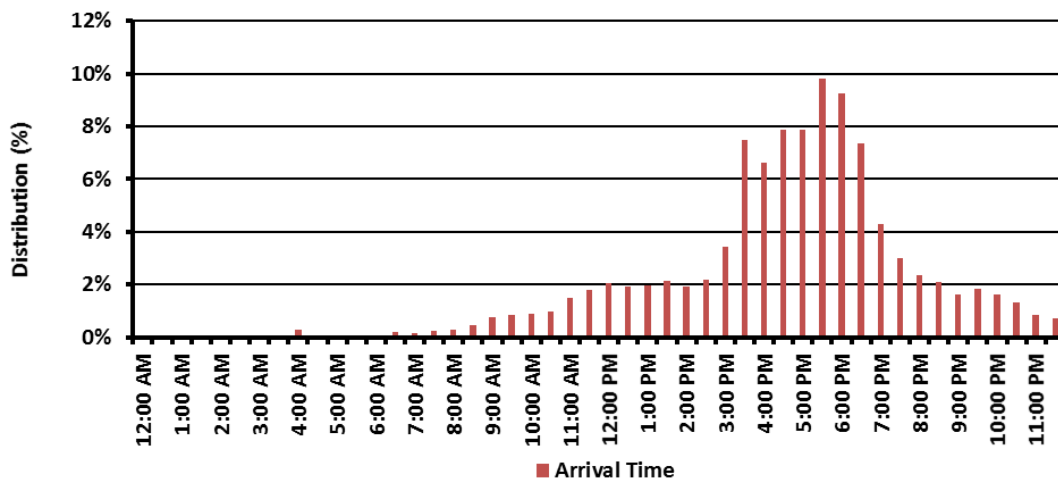
Source: Queensland Household Travel Survey

A.5.4 Destination Charging Completion Times

The charging completion time depends upon the start time, the assumed departure time, and the amount of charge required, which is in turn dependent on the daily driving distance. Generally speaking, the charging management function attempts to recharge the vehicle as quickly as possible while maximising the impact on minimum demand and minimising the impact on maximum demand.

The model uses the departure time distribution shown in Figure 52.

Figure 52 – Vehicle Departure Distribution by Vehicle Type



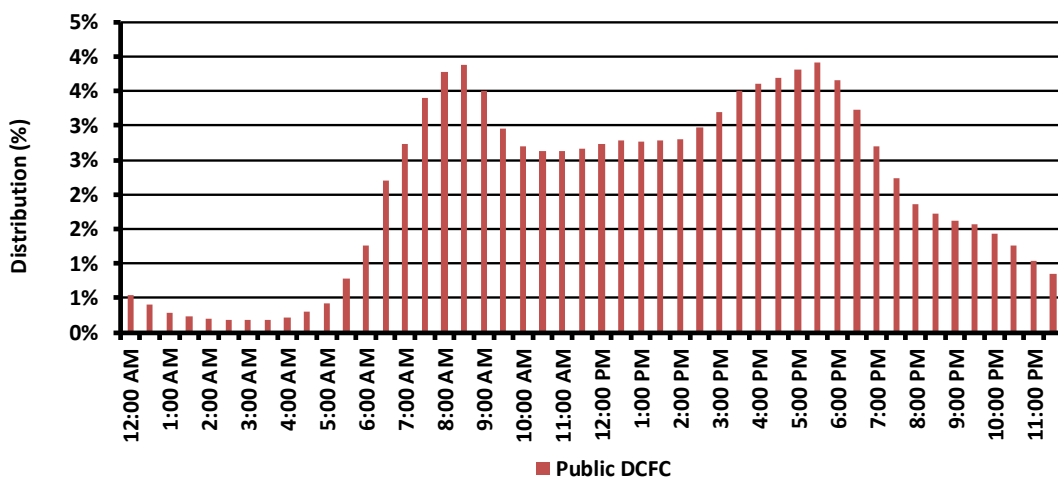
Source: Queensland Household Travel Survey

A.5.5 DCFC Charging Times

EV fast charging starts as soon as the vehicle arrives at the charging station and is completed within 5 minutes using 1MW chargers by 2036.

The charging start time is based on the Victorian Managing Traffic Congestion report and uses the traffic volume by time of day to determine the distribution of DCFC use, this is shown in Figure 53.

Figure 53 – Arrival Time Distribution



Source: Victorian Managing Tariff Congestion Report

A.6 Vehicle Stock Model

The vehicle stock model uses the following approach to determine overall change in stock for each state.

$$ICE_t = \sum_{i,j} \left[ICE_{i,j(t-1)} + (Vehicle\ Sales_{i,j(t)} - EV\ Uptake_{i,j(t)}) - \text{if} \left(t \leq AvgLifetime, \frac{ICE_{i,j(0)}}{AvgLifetime}, 0 \right) \right]$$

$$EV_t = \sum_{i,j} \left[EV_{i,j(t-1)} + EV\ Uptake_{i,j(t)} - \text{if} \left(t \leq AvgLifetime, \frac{EV_{i,j(0)}}{AvgLifetime}, 0 \right) \right]$$

Where:

- ICE_t = Total stock of ICE vehicles in year t
- EV_t = Total stock of EV vehicles in year t
- ICE_0 = Opening stock of ICE vehicles
- EV_0 = Opening stock of EV vehicles
- $ICE_{i,j(t-1)}$ = Stock of ICE vehicles in market i in class j in year t-1
- $EV_{i,j(t-1)}$ = Stock of EV vehicles in market i in class j in year t-1
- $EV\ Uptake_{i,j(t)}$ = % EV sales in market i in class j in year t
- $Vehicle\ Sales_{i,j(t)}$ = Vehicle sales in market i in class j in year t
- Average Lifetime = Average vehicle lifetime

A.6.1 Opening Stock

The opening stock of vehicles by vehicle class is sourced from VFACTS data for the calendar year 2016⁹ for EV and ICE vehicles by state. The opening stock feeds into the vehicle stock model at t=0 in the above equations.

A.6.2 Market Growth

Each year, each vehicle class in their respective market is assumed to grow at a constant rate per capita based on ABS forecasts of low, neutral and high population growth.

A.6.2 Average Lifetime

Average vehicle lifetime of all ICE vehicles is assumed to be 22 years based on ABS data¹⁰, while the average vehicle lifetime of all EVs are assumed to be 10 years.

⁹ Federal Chamber of Automotive Industry, VFACTS December National Report, 2016

¹⁰ ABS 9208.0 - Survey of Motor Vehicle Use, Australia, 12 months ended 30 June 2016

A.7 Policy Settings

A.7.1 Fuel Efficiency Standards

The Government's proposed fuel efficiency standards will improve the fuel efficiency of Australia's light vehicle fleet and bring Australia into line with international standards reducing greenhouse gas emissions from all light vehicles from the current 192gCO₂/km to 105gCO₂/km. That standard must be met across Australia's light vehicle fleet rather than on an individual vehicle model basis.

The proposed policy is expected to increase the average upfront cost of an ICE vehicle but reduce their average fuel expenditure over time. EVs are assumed to generate zero CO₂, given their fuel is already subject to the CET or a comparable CO₂ mechanism, enabling OEM's to reduce their compliance burden by selling EVs. This mechanism is expected to lead to OEM based cross-subsidisation of EVs up to the equivalent ICE cost, which the most recent analysis has found to average \$1,500.¹¹

The key vehicle policy assumptions and their impact on Energeia's modelling inputs are shown in Table 13 by sensitivity. The higher than \$1,500 assumptions are based on the potential for an EV multiplier to increase the value of EV based compliance. They also allow for the potential for additional, state based incentives.

Table 13 – Fuel Efficiency Standards Key Assumptions

Assumption	Weak Sensitivity	Neutral Sensitivity	Strong Sensitivity
Standard Introduction Date	2023	2021	2019
Impact on EV Incentives	\$1,500	\$2,500	\$5,000

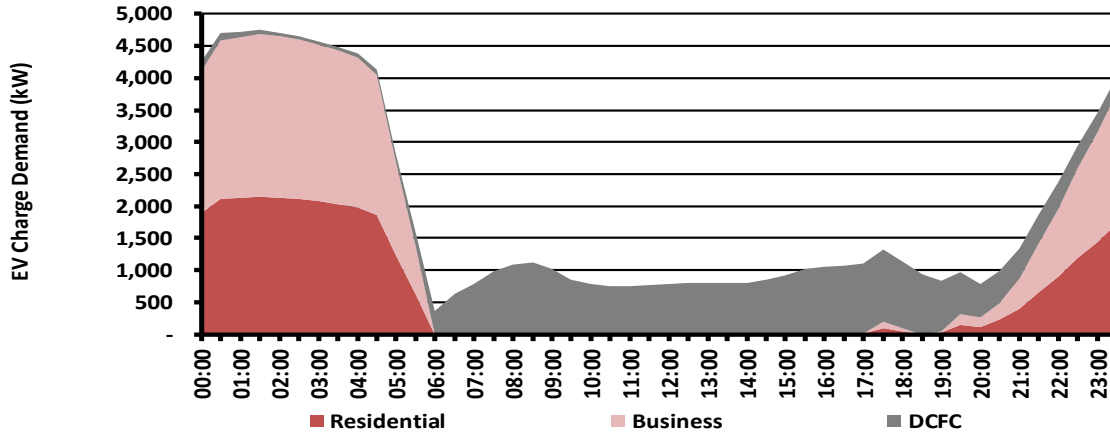
Source: Energeia, Climate Change Authority – Light Vehicle Emissions Standards Report 2014

¹¹ ,Climate Change Authority – Light Vehicle Emissions Standards Report 2014

B.4 EV Load Profiles

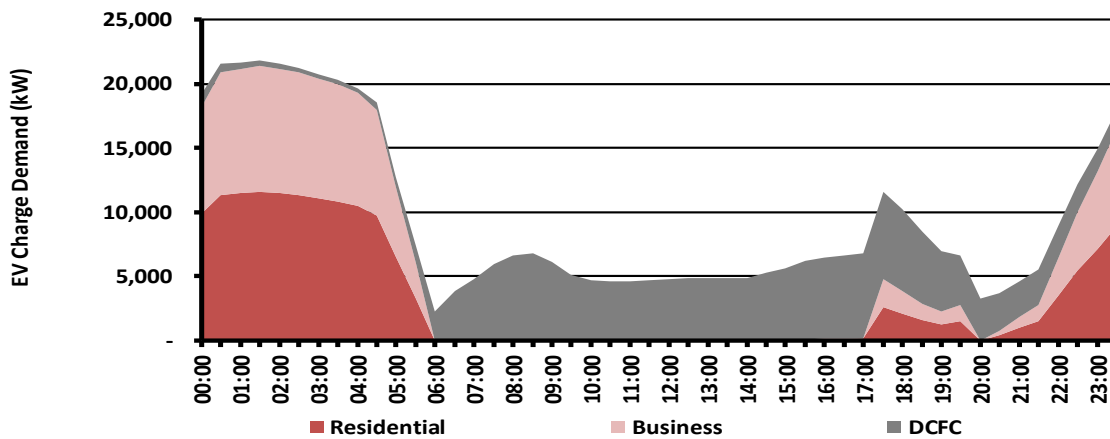
B.3.1 Queensland - Neutral

2020



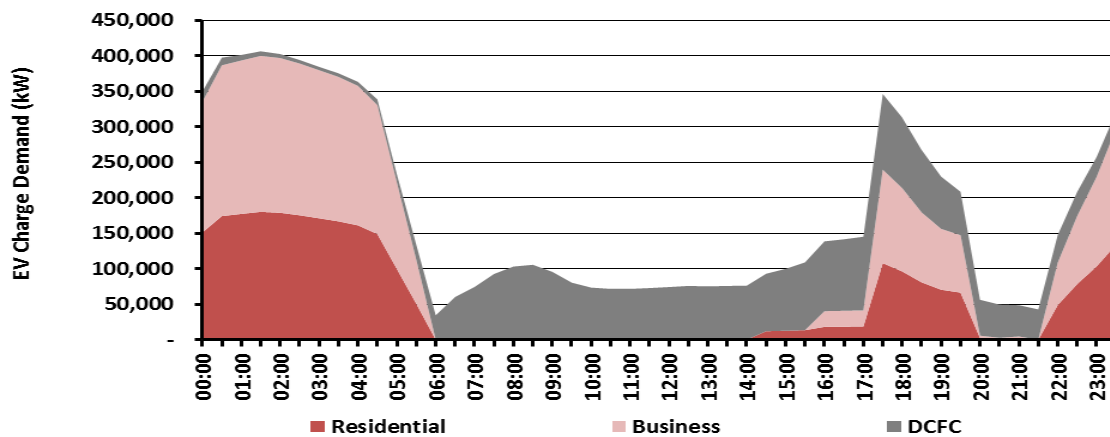
Source: Energeia

2025



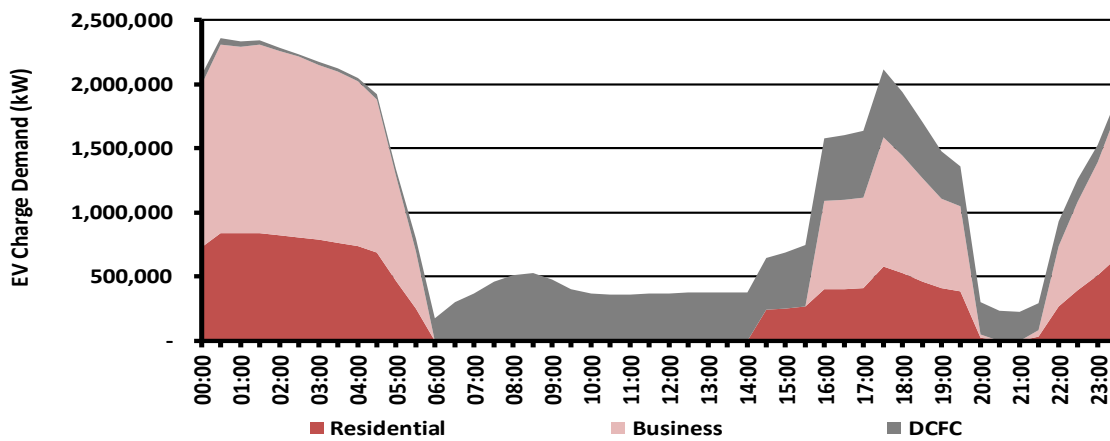
Source: Energeia

2036



Source: Energeia

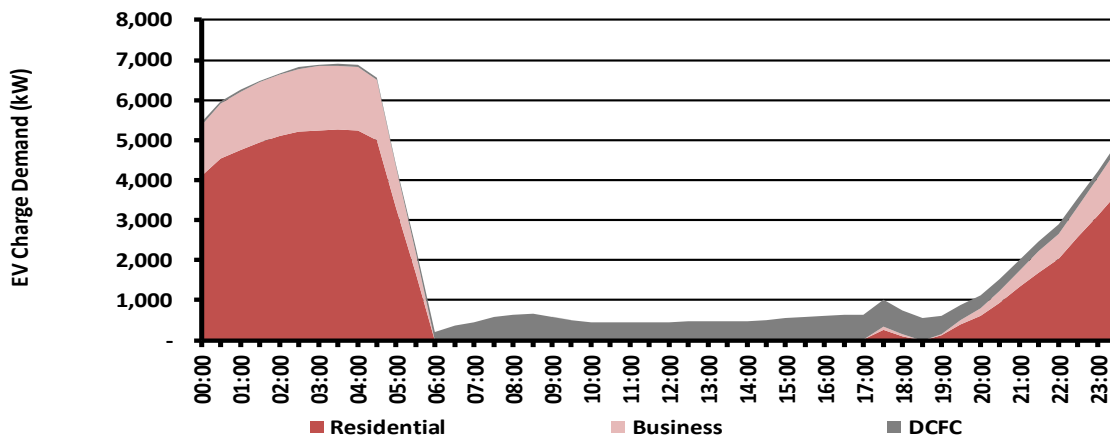
2050



Source: Energeia

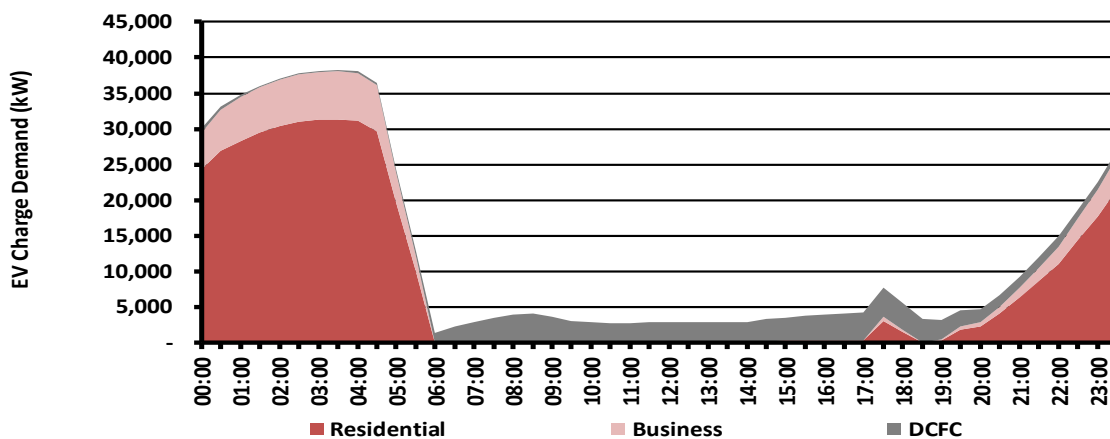
B.3.2 New South Wales - Neutral

2020



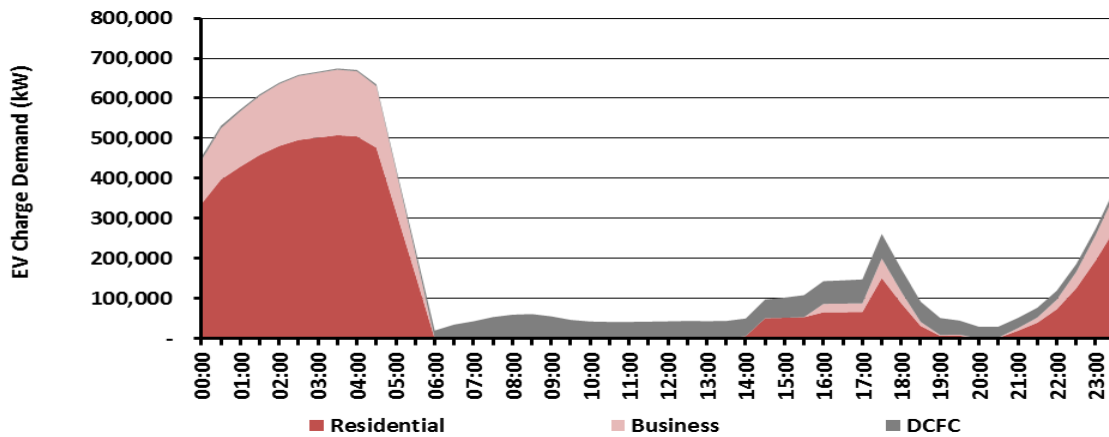
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2025



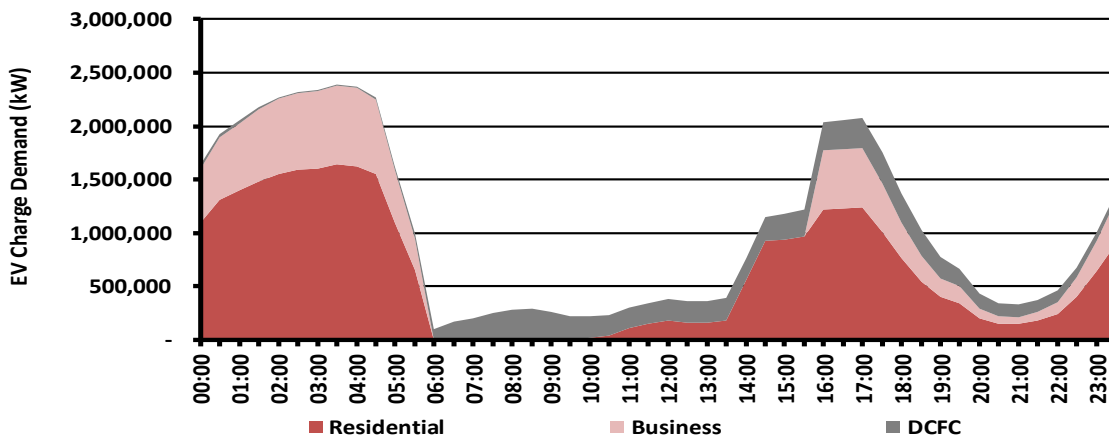
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2036



Source: Energeia

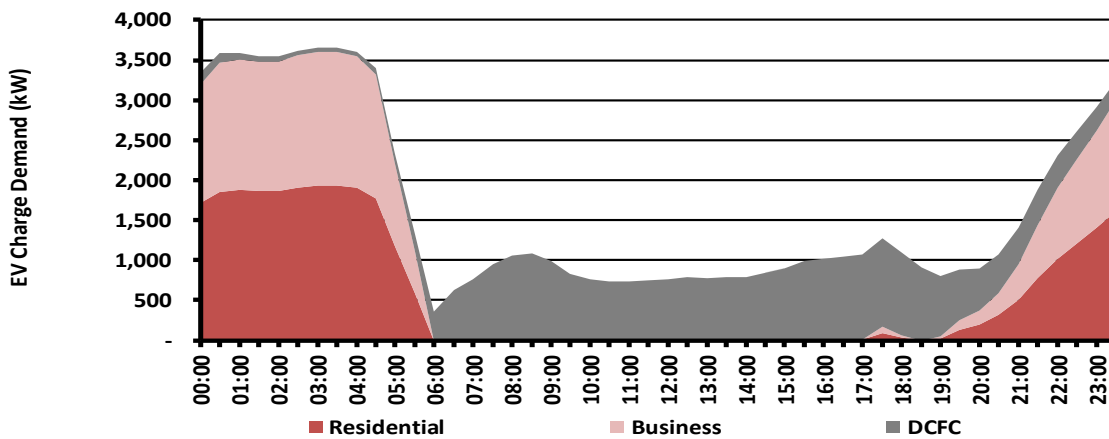
2050



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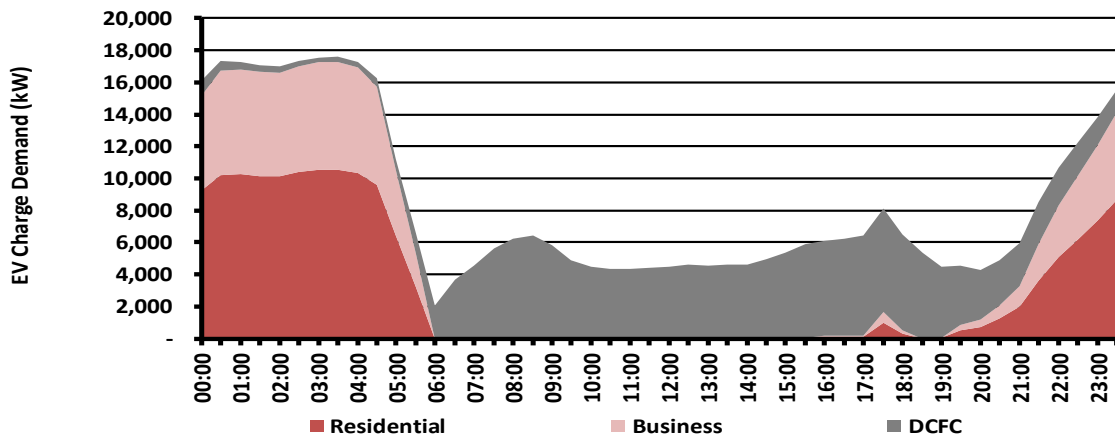
B.3.3 Victoria – Neutral

2020



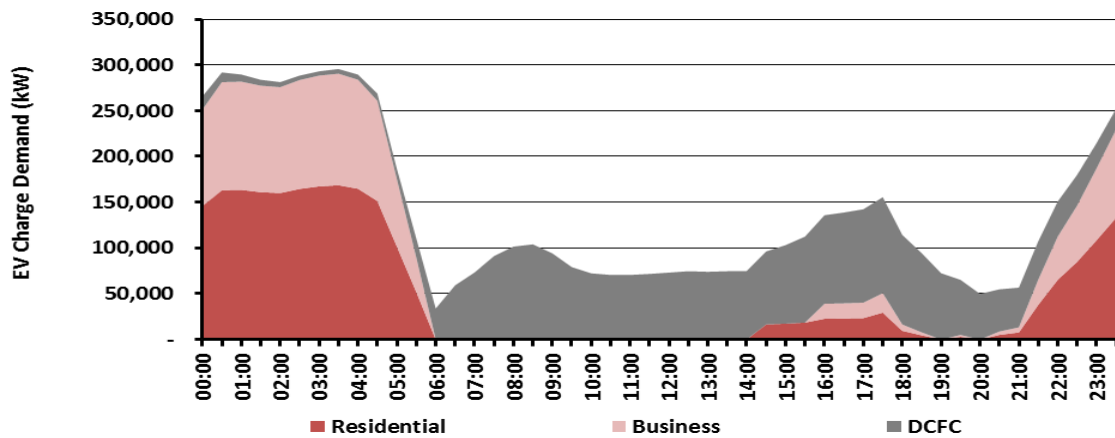
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2025



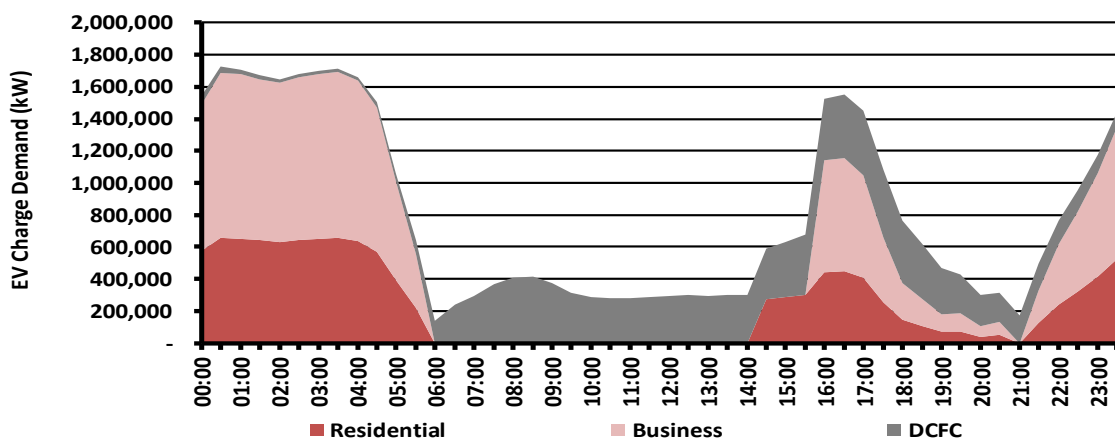
Source: Energeia

2036



Source: Energeia

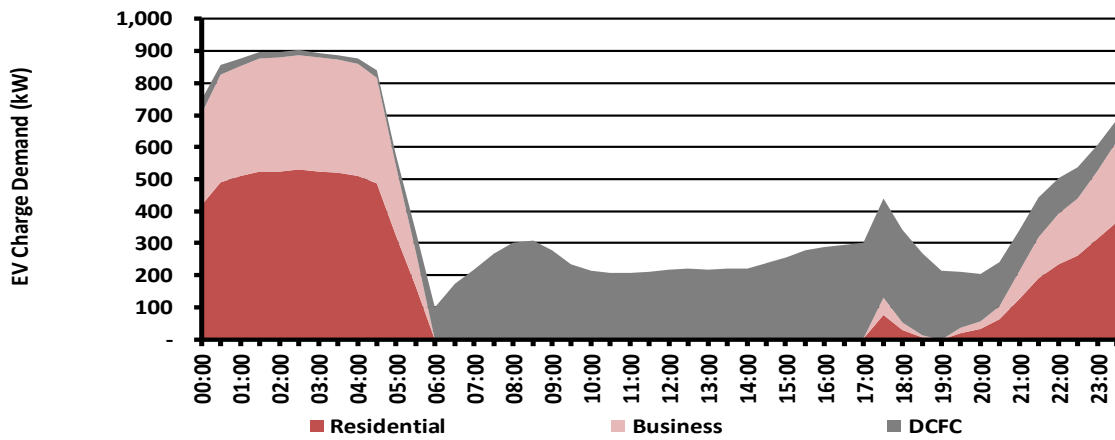
2050



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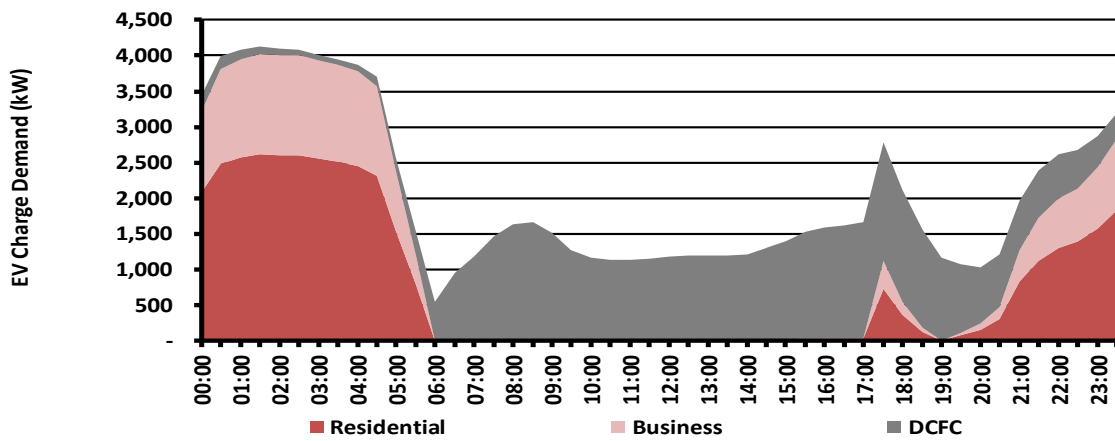
B.3.4 South Australia - Neutral

2020



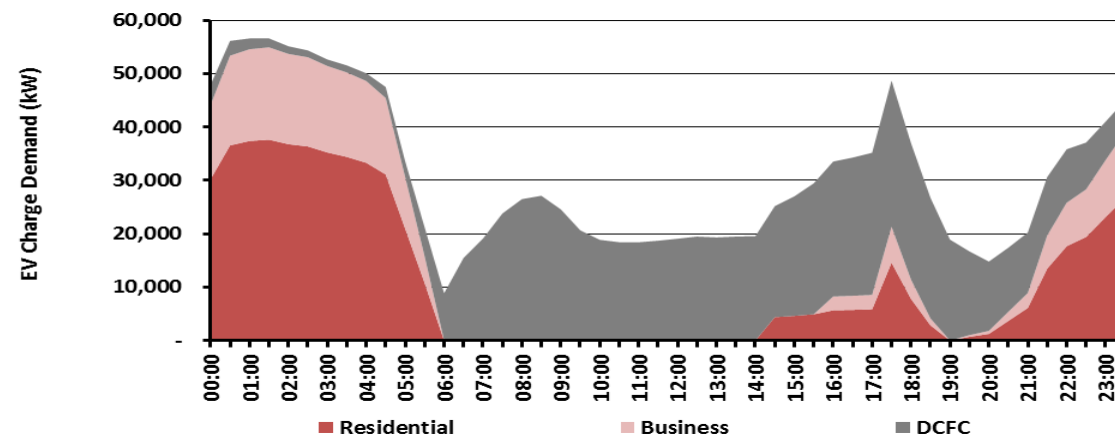
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2025



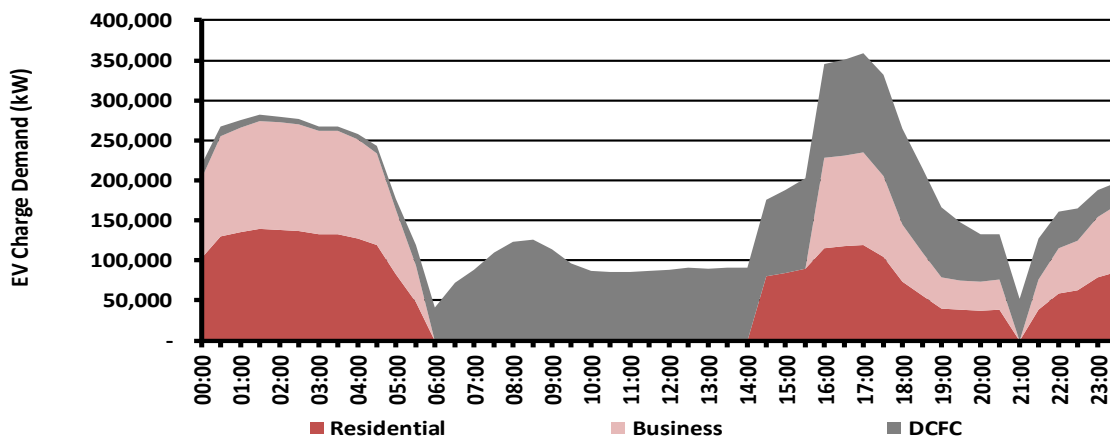
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2036



Source: Energeia

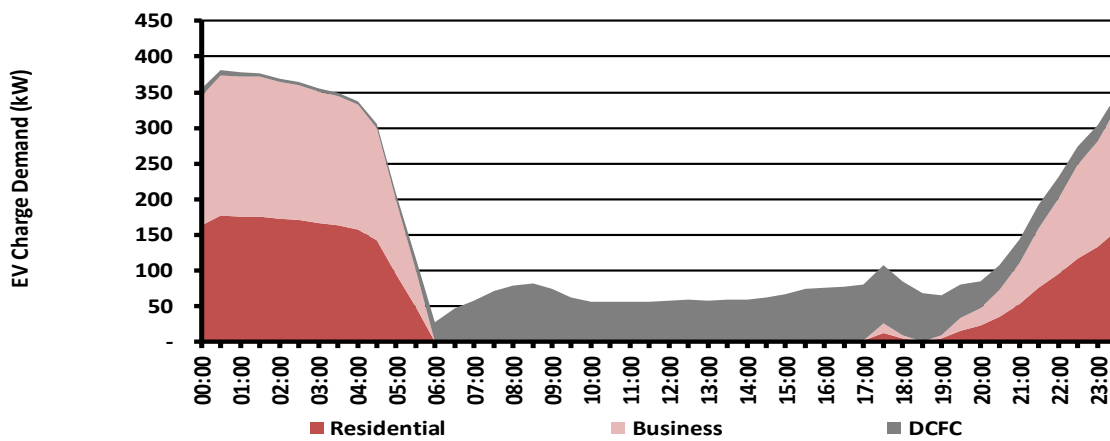
2050



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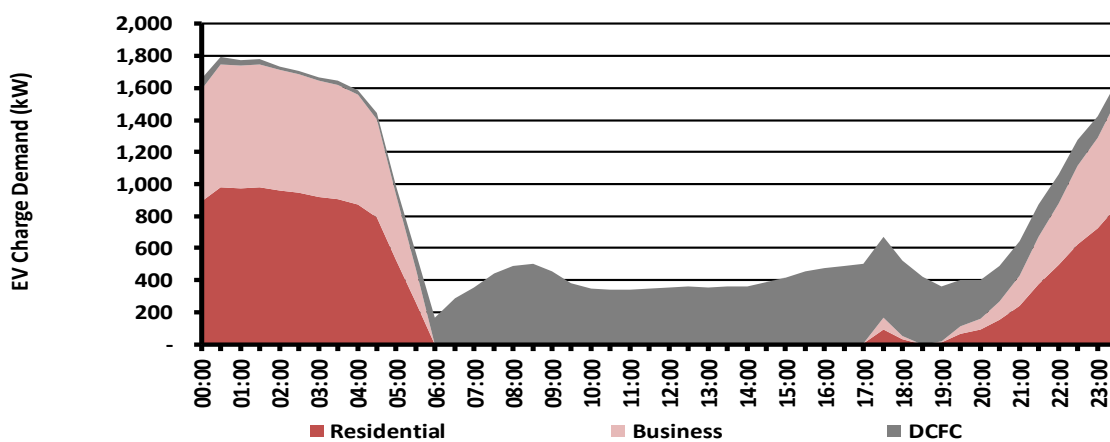
B.3.5 Tasmania - Neutral

2020



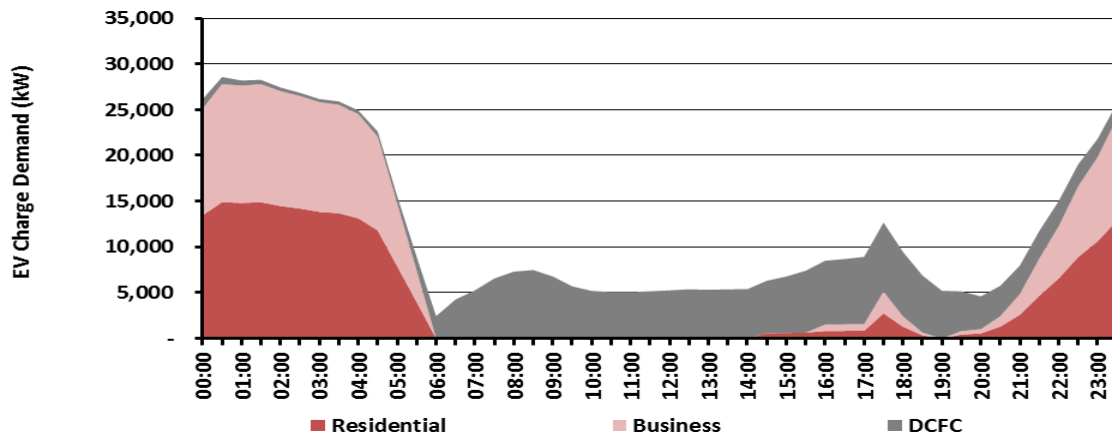
Source: Energeia

2025



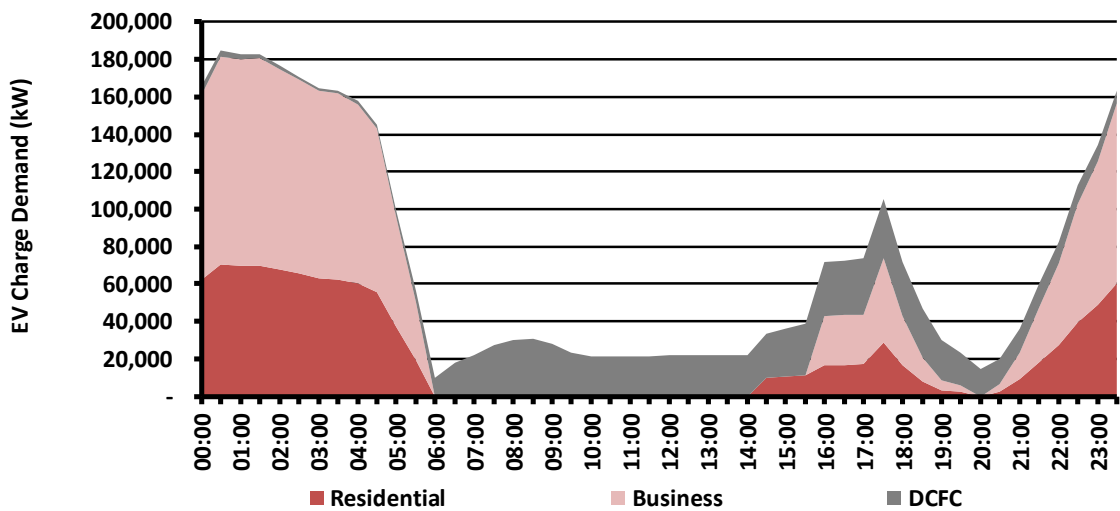
Source: Energeia

2036



Source: Energeia

2050



Source: Energeia