Appendix X:

2016 Greenhouse Gas Emissions Inventory and Projections for the San Diego Region

Final

December 2021

Table of Contents

Table of Figures

Table of Tables

Appendix X: 2016 Greenhouse Gas Emissions Inventory and Projections for the San Diego Region

Introduction

The San Diego Association of Governments (SANDAG) contracted the Energy Policy Initiatives Center (EPIC) at the University of San Diego (USD) to estimate the 2016 greenhouse gas (GHG) emissions for the San Diego region and to project GHG emissions for the years 2025, 2030, 2035, 2045, and 2050. The projections take into account the effect of existing federal and California (state) regulations and regional policies to reduce GHG emissions. GHG emissions estimates and projections are to be included in San Diego Forward: The 2021 Regional Plan (2021 Regional Plan) and its associated Environmental Impact Report (EIR). This appendix summarizes the methodologies and data used to conduct this analysis.

To the extent possible, EPIC followed the same methods used in developing the 2012 GHG emissions inventory and projections for San Diego Forward: The 2015 Regional Plan.[1] The 2016 GHG inventory and projections include 15 categories of emissions calculated based on the U.S. Community Protocol for Accounting and Reporting of Greenhouse Gas Emissions and California Air Resources Board (CARB) California statewide GHG inventory methodology.

Overview of the Appendix

This appendix includes the following sections:

- **Background** provides common background sources and assumptions used for the inventory and projections.
- **Summary of Results** provides the results of the 2016 GHG inventory and the GHG projections.
- **Method to Calculate Emissions Inventory and Projections by Category** includes subsections that cover the methods used to develop the inventory and projections by emissions category. Each subsection also describes how the methods to calculate the 2016 GHG inventory may vary from those used in the previous 2012 GHG inventory.

1 SANDAG: San Diego Forward: 2015 Regional Plan (2015).

Background

Greenhouse Gases

The primary GHGs included in this document are carbon dioxide (CO_2), methane (CH_4), and nitrous oxide (N_2O); others are included where data is available. Each GHG has a different capacity to trap heat in the atmosphere, known as its global warming potential (GWP), which is normalized relative to CO_2 and expressed in carbon dioxide equivalents (CO_2e). The 100-year GWPs reported by the Intergovernmental Panel on Climate Change (IPCC) are used by CARB to estimate GHG emissions inventories statewide.[2] The GWPs in this document, provided in Table X.1, are from the IPCC Fourth Assessment Report (AR4).[3]

Table X.1: Global Warming Potentials Used in the Regional Greenhouse Gas Inventory and Projections

Global Warming Potentials Used in the Regional Greenhouse Gas Inventory and Projections

Greenhouse Gas	Global Warming Potential
Carbon dioxide (CO_2)	1
Methane (CH_4)	25
Nitrous oxide (N_2O)	298
Difluoromethane (HFC-32)	675
1,1,1,2-Tetrafluoroethane (HFC-134a)	1,430
Pentafluoroethane (HFC-125)	3,500
1,1,1-Trifluoroethane (HFC-143a)	4,470
Carbon tetrafluoride (CF_4)	7,390
Octafluoropropane (C_3F_8)	8,830
1,1,1,3,3,3-Hexafluoropropane (HFC – 236fa)	9,810
Octafluorocyclobutane (C_4F_8)	10,300
Hexafluoroethane (C_2F_6)	12,200
Fluoroform (HFC-23)	14,800
Nitrogen trifluoride (NF_3)	17,200
Sulfur hexafluoride (SF_6)	22,800

Source: IPCC 2013.

2 CARB: Current California GHG Emission Inventory Data. 2000–2018 GHG Inventory (2020 Edition).

3 IPCC Fourth Assessment Report: Climate Change 2007: Direct Global Warming Potentials (2013).

Demographics

SANDAG estimates and forecasts population, housing, and employment for the San Diego region. The demographic estimates and projections through 2050 are provided in Table X.2.[4]

Table X.2: Demographic Estimates and Projections in the San Diego Region

Demographic Estimates and Projections in the San Diego Region				
Year	**Population**	**Jobs**	**Manufacturing Jobs***	**Housing Units**
2016	3,287,280	1,646,419	109,234	1,182,983
2025	3,470,848	1,761,747	116,046	1,288,216
2030	3,552,485	1,842,250	121,359	1,351,366
2035	3,620,348	1,921,475	126,618	1,409,866
2045	3,719,685	2,044,625	134,848	1,460,855
2050	3,746,073	2,086,318	137,503	1,471,299

* Manufacturing jobs are included in jobs.

2016 population and housing data are estimates. The rest are projections based on the SANDAG Series 14 Regional Growth Forecast (2021 Regional Plan).

Source: SANDAG 2020, 2021.

Rounding of Values in Tables and Figures

Rounding is used only for the final GHG values within the tables and figures throughout the document. Values are rounded to the nearest integer of a higher order of magnitude. Values are not rounded in the intermediary steps in the actual calculation. Because of rounding, some totals may not equal the exact values summed in any table or figure.

Summary of Results

Table X.3 provides a summary of the 2016 GHG inventory and the GHG projections in the San Diego region.

[4] 2016 population and housing are from the SANDAG Demographic & Socio-Economic Estimates (August 19, 2020, Version). SANDAG Data Surfer, accessed on December 10, 2020. Other estimates and projections are based on SANDAG Series 14 Growth Forecast, provided by SANDAG staff to EPIC, March 29, 2021.

Table X.3: Summary of 2016 Greenhouse Gas Inventory and Greenhouse Gas Projections

Summary of 2016 Greenhouse Gas Inventory and Greenhouse Gas Projections						
Greenhouse Gas Emissions (Million Metric Tons [MMT] CO_2e)						
Emissions Category	**2016**	**2025**	**2030**	**2035**	**2045**	**2050**
Passenger Cars and Light-Duty Vehicles* (No SAFE Rule Impact)	10.4	8.0 (7.8)	7.2 (6.8)	6.4 (5.8)	6.3 (5.6)	6.3 (5.6)
Electricity	5.3	3.4	1.9	1.3	0.2	0.2
Natural Gas	3.1	3.3	3.4	3.4	3.5	3.6
Industrial	2.1	2.2	2.3	2.4	2.5	2.5
Heavy-Duty Trucks and Vehicles	1.8	1.7	1.7	1.7	1.7	1.7
Other Fuels	1.1	1.4	1.4	1.5	1.5	1.5
Off-Road Transportation	0.62	0.72	0.79	0.83	0.91	0.95
Solid Waste	0.59	0.62	0.64	0.65	0.67	0.67
Water	0.24	0.28	0.22	0.15	—	—
Aviation	0.21	0.29	0.32	0.34	0.40	0.43
Rail	0.11	0.17	0.18	0.19	0.20	0.20
Wastewater	0.07	0.08	0.08	0.08	0.08	0.08
Agriculture	0.05	0.06	0.06	0.06	0.06	0.06
Marine Vessels	0.05	0.06	0.06	0.06	0.08	0.08
Soil Management	0.05	0.04	0.04	0.04	0.04	0.04
Total* (Total: No SAFE Rule Impact)	**26**	**22 (22)**	**20 (20)**	**19 (18)**	**18 (17)**	**18 (18)**

* Includes GHG impact of SAFE Rule.

MMT – million metric tons.

SAFE Rule – Federal Safer Affordable Fuel-Efficiency Vehicles Rule, April 2020.

2016 is an inventory year, and the rest are forecast years. The GHG emissions projections include the impact of federal and state regulations and regional policies and programs to reduce GHG emissions.

Source: EPIC, USD 2021.

In September 2019, the U.S. Environmental Protection Agency (EPA) and the National Highway Traffic Safety Administration (NHTSA) published the "Safer Affordable Fuel-Efficiency (SAFE) Vehicles Rule Part One: One National Program" (SAFE Rule Part One). The SAFE Rule Part One revoked California's authority to set its own GHG emissions standards and set zero-emissions vehicle (ZEV) mandates. In April 2020, EPA and NHTSA issued the Final SAFE Rule that relaxed federal GHG emissions and Corporate Average Fuel Economy (CAFE) standards for model year 2021–2026 vehicles. The GHG emissions from passenger car and light-duty vehicles and total GHG emissions with and without the SAFE rule impact are shown in Table X.3. The method to adjust on-road transportation emissions with SAFE Rule is discussed in the section *On-Road Transportation – Passenger Car and Light-Duty Vehicles*.

The previous 2012 GHG inventory included the following land use and development influences on the regional inventory: (1) carbon sequestration from vegetation cover, (2) carbon emissions from vegetation displaced by development, and (3) carbon emissions from vegetation burning due to wildfires. This inventory excludes emissions and sequestration estimates from vegetation and follows CARB's approach to track statewide GHG emissions from anthropogenic activities not including the GHG flux associated with carbon stocks in California's natural and working lands[5] and wildfire emissions. This is because wildfires are part of Earth's carbon cycle and it is difficult to determine how much of the wildfire emissions are from anthropogenic activities.[6, 7]

The forecast includes the regional effects of existing federal and state polices and regulations to reduce GHG emissions. The projected reductions are based on the current implementation timeline of these regulations. Many regulations do not extend beyond 2025 or 2030, and therefore are assumed to have no additional impact after 2025 or 2030.

Method to Calculate Emissions Inventory and Projections by Category

On-Road Transportation – Passenger Car and Light-Duty Vehicles

The passenger car and light-duty vehicles emissions category is the largest contributor of GHG emissions in the San Diego region, accounting for about 40% of total GHG emissions in the 2016 inventory and 32% of total GHG emissions in the 2050 projection. Tailpipe GHG emissions from on-road transportation are the result of fuel combustion (i.e., gasoline, diesel, natural gas) from mobile vehicles on freeways, highways, and local roads. The vehicle classes included in this emissions category are passenger cars and light-duty vehicles. The GHG emissions from other on-road vehicles are accounted for in the subsection titled *On-Road Transportation – Heavy-Duty Trucks and Vehicles*.

Method Used to Estimate 2016 Emissions

EPIC used EMFAC2017, CARB's on-road mobile sources model, to estimate the on-road transportation emissions for passenger cars and light-duty vehicles.[8] SANDAG provided the input file to run EMFAC2017 under custom mode, as well as the output file containing all emissions results.[9] The input file, from SANDAG's activity-based model (ABM14.2.2), includes vehicle miles traveled (VMT) on an average weekday by EMFAC vehicle categories and fuel types. The output file, from an EMFAC2017 custom model run, provides CO_2 emissions in tons per weekday for each vehicle category and each fuel type.

5 CARB began a natural and working lands carbon and GHG flux assessment in 2018 based on IPCC principles. See arb.ca.gov/nwl-inventory.

6 CARB: Frequently Asked Questions: Wildfire Emissions.

7 California Senate Bill 901 (Dodd, 2018) (SB 901) requires that the state develops a report assessing GHG emissions from wildfire and forest management activities by December 2020 and every five years thereafter. The SB 901 2020 report provides wildfire estimates for the years 2000–2019. See California Wildfire Burn Acreages and Preliminary Emissions Estimates.

8 CARB: Mobile Source Emissions Inventory. EMFAC2017.

9 Files provided by SANDAG staff, October 6, 2021.

This passenger car and light-duty vehicles emissions category covers the GHG emissions from EMFAC2017 vehicle classes LDA, LDT1, LDT2, and MDV.[10]

To convert the emissions output from tons of CO_2 per weekday to metric tons of CO_2e per year, EPIC used the weekday-to-year conversion factor and CO_2-to-CO_2e (CO_2, CH_4, and N_2O) conversion factor for each EMFAC vehicle category, based on statewide GHG inventory assumptions and EMFAC2017 default run results, respectively.[11] The weekday-to-annual conversion factors for LDA, LDT1, LDT2 and MDV are all 347 weekdays per year; the CO_2 to CO_2e conversion factors range from 1.01 for gasoline LDT2 to 1.05 for diesel LDA.[12] The key inputs and results are shown in Table X.4.

Table X.4: Key Inputs and 2016 Greenhouse Gas Emissions from On-Road Transportation – Passenger Car and Light-Duty Vehicles

Key Inputs and 2016 Greenhouse Gas Emissions from On-Road Transportation – Passenger Car and Light-Duty Vehicles	
VMT (miles per weekday)*	78,987,431
CO_2 Emissions (tons per weekday)**	32,605
Conversion Factor (tons CO_2 per weekday to MT CO_2e per year)	319
GHG Emissions (MT CO_2e)	10,404,317
GHG Emissions (MMT CO_2e)	**10.4**

* SANDAG ABM14.2.2 VMT.

** EMFAC2017 model run with custom VMT inputs from ABM14.2.2.

Passenger car and light-duty vehicles are EMFAC2017 vehicle classes LDA, LDT1, LDT2, and MDV.

Source: CARB 2016, 2017; SANDAG 2021; EPIC, USD 2021.

Difference from Previous 2012 Inventory

Methods to estimate emissions from passenger car and light-duty vehicles are the same in both 2012 and 2016 regional GHG inventories. However, the previous version of SANDAG's ABM (ABM1) and the previous version of EMFAC (EMFAC2014) were used to calculate the 2012 GHG emissions. ABM2+ is used for analysis related to the 2021 Regional Plan (additional information included in Appendix S: Travel Demand Modeling Tools), and EMFAC2017 described below includes new regulations that were not reflected in EMFAC2014.

10 LDA: passenger cars; LDT1: light-duty trucks with gross vehicle weight rating (GVWR) less than 6,000 lbs and equivalent test weight (ETW) no larger than 3,750 lbs; LDT2: light-duty trucks with GVWR less than 6,000 lbs and ETW between 3,750 and 5,750 lbs; and MDV: medium-duty trucks with GVWR between 6,000 and 8,500 lbs.

11 This approach is recommended by CARB EMFAC staff. Personal communication, January 27, 2020.

12 The weekday-to-year conversion factors are based on CARB's California's 2004–2014 Greenhouse Gas Emission Inventory Technical Support Document, 2016 Edition, accessed March 23, 2020. The CO_2-to-CO_2e conversion factors are based on EMFAC2017 default 2016 emissions run for San Diego region by vehicle category and fuel type, January 14, 2020, model run.

Method Used to Develop Emissions Projections

The method used to develop projections is similar to the method used to estimate 2016 emissions, based on an EMFAC2017 model run with SANDAG VMT inputs. For forecast years, EMFAC2017 model results include the effect of all key federal and state laws, regulations, and legislative actions that were adopted as of December 2017. The updated regulation for passenger cars and light-duty vehicles since the release of EMFAC2014 is the California Advanced Clean Car (ACC) Program, which includes:

- Tailpipe emissions standards equivalent to CAFE standards for vehicle model years 2017–2025.
- A ZEV program that requires manufacturers to produce increasing numbers of ZEVs and plug-in hybrid electric vehicles for model years 2017–2025.

The impact of the ACC Program had already been incorporated into the previous version, EMFAC2014; however, EMFAC2017 includes updated assumptions in the ACC regulation based on its 2017 midterm review.

With the same tons of CO_2 per weekday to MT CO_2e per year conversion method discussed in the previous inventory method section, the key inputs and results are shown in Table X.5.[13]

13 VMT input files and emission output files were provided by SANDAG Staff, October 6, 2021.

Table X.5: Projected Greenhouse Gas Emissions from On-Road Transportation – Passenger Car and Light-Duty Vehicles after Federal and State Regulations

Projected Greenhouse Gas Emissions from On-Road Transportation – Passenger Car and Light-Duty Vehicles after Federal and State Regulations					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
VMT (miles per weekday)*	79,500,510	79,869,787	79,707,577	81,537,871	81,796,296
CO_2 Emissions (tons per weekday)**	24,510	21,321	19,510	18,916	18,738
Conversion Factor (tons CO_2 per weekday to MT CO_2e per year)***	318	318	317	317	318
GHG Emissions (MT CO_2e)	7,786,162	6,769,475	6,193,868	6,006,111	5,949,948
GHG Emissions (MMT CO_2e)	7.8	6.8	6.2	6.0	5.9

* 2025, 2030, 2035, and 2050 VMT direct outputs of SANDAG ABM14.2.2, 2045 VMT is interpolated linearly between 2040 and 2050 VMT.

** EMFAC2017 model run with custom VMT inputs from SANDAG ABM14.2.2, 2045 CO_2 emissions are interpolated linearly between 2040 and 2050.

*** Conversion factors vary slightly by year.

Source: CARB 2016, 2017; SANDAG 2021; EPIC, USD 2021.

The VMT projected under ABM do not capture the miles and trips avoided as a result of the following SANDAG off-model strategies in the 2021 Regional Plan:

- Electric vehicle (EV) strategies:
  - EV charger program
  - EV incentive program
- Shared mobility strategies:
  - Vanpool
  - Carshare
  - Pooled ride
  - Regional transportation demand management ordinance

The detailed strategy descriptions and the methods to estimate CO_2 reductions due to the strategies are provided in Appendix S: Travel Demand Modeling Tools. EPIC converted the annual CO_2 reductions (EV strategies) and weekday CO_2 reductions (shared mobility strategies) to annual CO_2e reductions using the same conversion method as described above. For the shared mobility strategies, only the GHG reductions from running exhaust and start exhaust processes are included in this appendix to be consistent with Appendix S. The projected GHG reductions from EV strategies and shared mobility strategies are shown in Table X.6 and Table X.7, respectively.

Table X.6: Projected Greenhouse Gas Reductions from SANDAG Electric Vehicle Off-Model Strategies

Projected Greenhouse Gas Reductions from SANDAG Electric Vehicle Off-Model Strategies		
Projection Year	**2035**	**2050**
GHG Reduction from EV Strategies: Regional Charger Program (MT CO_2 per year)*	103,617	267,677
GHG Reduction from EV Strategies: Vehicle Incentive Program (MT CO_2 per year)*	230,939	—
GHG Reduction from EV Strategies: Combined EV Charger and EV Incentive Programs (MT CO_2 per year)*	334,556	276,677
Conversion Factor (MT CO_2e per MT CO_2)**	1.01	1.01
GHG Reduction from SANDAG EV Strategies (MT CO_2e)	337,252	278,990
GHG Reduction from SANDAG EV Strategies (MMT CO_2e)	0.34	0.28

* GHG reduction from the programs and program design are described in 2021 Regional Plan Appendix S. Because off-model strategies are intended for use in complying with Senate Bill 375 (Steinberg, 2008) GHG emissions reduction targets, 2035 is the primary year of analysis and reductions associated with interim years are not provided.

** EMFAC2017 assumptions for passenger car and light-duty vehicle classes: LDA, LDT1, LDT2, and MDV.

Source: CARB 2017; SANDAG 2021; EPIC, USD 2021.

Table X.7: Projected Greenhouse Gas Reductions from SANDAG Shared Mobility Strategies

Projected Greenhouse Gas Reductions from SANDAG Shared Mobility Strategies		
Projection Year	**2035**	**2050**
Vehicle Trips Avoided		
Vanpool Strategy (trips avoided per weekday)*	7,853	8,837
Pooled Rides Strategy (trips avoided per weekday)*	2,123	2,106
Transportation Demand Management Ordinance Strategy (trips avoided per weekday)*	46,121	72,436
Total (trips avoided per weekday)	56,097	83,379
Total (trips avoided per year)**	19,465,777	28,932,559
GHG Emissions per Trip Start (grams CO_2e per trip)***	46	42
GHG Reduction due to Trips Avoided (MT CO_2e)	887	1,209
Vehicle Miles Avoided		
Vanpool Strategy (miles avoided per weekday)*	400,805	450,486
Carshare (miles avoided per weekday)*	176,896	—
Pooled Rides Strategy (miles avoided per weekday)*	11,658	11,540
Transportation Demand Management Ordinance Strategy (miles avoided per weekday)*	364,337	579,172
Total (miles avoided per weekday)	953,696	1,041,198
Total (miles avoided per year)**	330,932,552	361,295,803
GHG Emissions per Mile (grams CO_2e per mile)***	218	202
GHG Reduction due to Miles Avoided (MT CO_2e)	72,018	73,097
Total (Trips + Miles Avoided)		
GHG Reduction from Shared Mobility Strategies (MT CO_2e)	72,905	74,306
GHG Reduction from Shared Mobility Strategies (MMT CO_2e)	0.07	0.07

* GHG reduction from the programs and program design are described in 2021 Regional Plan Appendix S. The carshare strategy does not have trips avoided or miles avoided in 2050.

** 347 weekdays per year, EMFAC2017 assumptions for passenger car and light-duty vehicle classes: LDA, LDT1, LDT2, and MDV.

*** Based on the total number of trips, VMT, start exhaust (EMFAC2017 process STARTEX), and running exhaust (EMFAC2017 process RUNEX) CO_2e emissions from LDA, LDT1, LDT2, and MDV vehicle classes (EMFAC2017 model run with ABM14.2.2 inputs).

Source: CARB 2016, 2017; SANDAG 2021; EPIC, USD 2021.

The projected emissions from passenger car and light-duty vehicles after impacts of federal and state regulations and SANDAG programs are shown in Table X.8.

Table X.8: Projected Greenhouse Gas Emissions from On-Road Transportation – Passenger Car and Light-Duty Vehicles

Projected Greenhouse Gas Emissions from On-Road Transportation – Passenger Car and Light-Duty Vehicles					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
GHG Emissions after Federal and State Regulations (MMT CO_2e)	7.8	6.8	6.2	6.0	5.9
GHG Reduction from SANDAG Electric Vehicle Strategies* (MMT CO_2e)	N/A	N/A	−0.34	−0.30	−0.28
GHG Reduction from SANDAG Shared Mobility Strategies* (MMT CO_2e)	N/A	N/A	−0.07	−0.07	−0.07
GHG Emissions (MMT CO_2e)	**7.8**	**6.8**	**5.8**	**5.6**	**5.6**

* 2045 GHG reductions are interpolated linearly between 2035 and 2050 GHG reductions in Table X.6 and Table X.7.

Source: SANDAG 2021; EPIC, USD 2021.

In April 2020, EPA and NHTSA issued the Final SAFE Rule that relaxed federal GHG emissions and CAFE standards for model year 2021–2026 vehicles. In June 2020, CARB released off-model adjustment factors to adjust tailpipe CO_2 emissions outputs from EMFAC models to account for the impacts of the SAFE Rule. The adjustment factors are for gasoline light-duty vehicles (EMFAC2017 vehicle categories LDA, LDT1, LDT2, and MDV) only and in the form of multipliers applied to emissions outputs from the EMFAC model. The SAFE Rule adjustment factors and projection results are shown in Table X.9.[14]

[14] CARB: EMFAC Off-Model Adjustment Factors for Carbon Dioxide (CO_2) Emissions to Account for the SAFE Vehicles Rule Part One and the Final SAFE Rule (2020), accessed September 3, 2020. Method to apply adjustment factors were confirmed by CARB EMFAC staff. Personal communication between EPIC and CARB, June 30, 2020.

Table X.9: SAFE Rule Adjustment Factors and Projected Emissions: On-Road Transportation – Passenger Car and Light-Duty Vehicles

SAFE Rule Adjustment Factors and Projected Emissions: On-Road Transportation – Passenger Car and Light-Duty Vehicles			
EMFAC2017 Vehicle Category	**CO_2 Emissions (Tons per Weekday)***	**SAFE Rule Adjustment Factor****	**Adjusted CO_2 Emissions (Tons per Weekday)****
2025 Projection			
LDA – GAS	13,383	1.031	13,796
LDT1 – GAS	1,688	1.031	1,740
LDT2 – GAS	5,138	1.031	5,297
MDV – GAS	4,019	1.031	4,143
LDVs – NON-GAS	282	N/A	282
Total LDV	24,510	N/A	**25,259**
2030 Projection			
LDA – GAS	11,994	1.070	12,836
LDT1 – GAS	1,492	1.070	1,597
LDT2 – GAS	4,266	1.070	4,566
MDV – GAS	3,307	1.070	3,540
LDVs – NON-GAS	262	N/A	262
Total LDV	21,321	N/A	**22,799**
2035 Projection			
LDA – GAS	11,164	1.100	12,277
LDT1 – GAS	1,369	1.100	1,505
LDT2 – GAS	3,789	1.100	4,167
MDV – GAS	2,942	1.100	3,235
LDVs – NON-GAS	246	N/A	246
Total LDV	19,510	N/A	**21,431**
2050 Projection			
LDA – GAS	10,922	1.127	12,312
LDT1 – GAS	1,291	1.127	1,455
LDT2 – GAS	3,540	1.127	3,991
MDV – GAS	2,744	1.127	3,093
LDVs – NON-GAS	241	N/A	241
Total LDV	18,738	N/A	**21,091**

Source: CARB 2017, 2020; SANDAG 2021; EPIC, USD 2021.

Using the same conversion method for tons of CO_2 per weekday to MT CO_2e per year discussed in the inventory method section, the results are shown in Table X.10.

Table X.10: Projected GHG Emissions from On-Road Transportation – Passenger Car and Light-Duty Vehicles after Federal and State Regulations with SAFE Rule Impact

Projected GHG Emissions from On-Road Transportation – Passenger Car and Light-Duty Vehicles after Federal and State Regulations with SAFE Rule Impact

Projection Year	2025	2030	2035	2045	2050
VMT (miles per weekday)*	79,500,510	79,869,787	79,707,577	81,537,871	81,796,296
Adjusted CO_2 Emissions (tons per weekday)**	25,259	22,799	21,431	21,194	21,091
Conversion Factor (tons CO_2 per weekday to MT CO_2e per year)	318	317	317	317	318
Business as Usual (BAU) GHG Emissions (MT CO_2e)	8,021,827	7,234,851	6,803,306	6,728,821	6,696,674
BAU GHG Emissions (MMT CO_2e)	8.0	7.2	6.8	6.7	6.7
GHG Reduction from SANDAG EV Strategies (MMT CO_2e)***	N/A	N/A	−0.37	−0.34	−0.31
GHG Reduction from SANDAG Shared Mobility Strategies (MMT CO_2e)***	N/A	N/A	−0.08	−0.08	−0.08
GHG Emissions (MMT CO_2e)**	**8.0**	**7.2**	**6.4**	**6.3**	**6.3**

* 2025, 2030, 2035, and 2050 VMT direct outputs of SANDAG ABM14.2.2; 2045 VMT is interpolated linearly between 2040 and 2050 VMT.

** EMFAC2017 model run with custom VMT inputs from SANDAG adjusted with SAFE Rule impact, as shown in Table X.9; 2045 CO_2 adjusted emissions are interpolated linearly between 2040 and 2050 adjusted emissions.

*** GHG reductions from EV strategies (Table X.6) and from EV strategies (Table X.7) with SAFE Rule adjustment factors (Table X.9); 2045 GHG reductions are interpolated linearly between 2035 and 2050 GHG reductions.

**** GHG emissions after the impact of federal and state regulations (BAU emissions), SANDAG EV Strategies and SANDAG Shared Mobility Strategies.

Source: CARB 2017, 2020; SANDAG 2021; EPIC, USD 2021.

Electricity

GHG emissions from electricity use in the San Diego region account for 20% of total emissions in the 2016 inventory and 1% in the 2050 projection.

Method Used to Estimate 2016 Emissions

To estimate GHG emissions from grid-supply electricity use, EPIC adjusted the 2016 electricity sales with transmission and distribution losses and multiplied the adjusted sales by the electricity emission factor, expressed in pounds of CO_2e per megawatt-hour (lbs CO_2e/MWh).

The local utility, San Diego Gas & Electric (SDG&E), provided the 2016 San Diego regional electricity sales by bundled and Direct Access (DA) supply for each customer class. The San Diego regional electricity sales account for electricity sales to all local jurisdictions, including military bases and tribal reservations.[15] The transmission and distribution loss factor, 0.082, is the loss estimate for the entire SDG&E service territory (larger than the San Diego region) and accounts for the difference between electricity generated for load and electricity sales.[16]

SDG&E and electric service providers (ESPs) for DA customers have different power mixes in their electricity supplies. The SDG&E 2016 bundled emission factor, 527 lbs CO_2e/MWh, was calculated using Federal Energy Regulatory Commission Form 1 data, the California Energy Commission (CEC) Power Source Disclosure Program data on SDG&E-owned and purchased power, and EPA's Emissions and Generating Resource Integrated Database (eGRID) on specific power plant emissions. EPIC's technical working paper, "Estimating Annual Average Greenhouse Gas Emission Factors for the Electricity Sector: A Method for Inventories," describes the detailed method to calculate the SDG&E bundled electricity emission factor.[17] The DA emission factor, 836 lbs CO_2e/MWh, is a default taken from the California Public Utilities Commission Decision 14-12-037.[18]

Two adjustments are made to the emissions estimate based on grid-supply electricity:

- Emissions associated with electricity use at water treatment plants in the San Diego region were allocated to the water category and removed from the electricity category. The method used to identify electricity use at water treatment plants is discussed in the *Water* section of this appendix.
- Emissions associated with natural gas used for on-site self-serve electric generation, mostly attributed to co-generation plants, were removed from the natural gas category and allocated to the electricity category. EPIC used the CEC Quarterly Fuel and Energy Report (QFER) Power Plant Owner Reporting database, U.S. Energy

15 Electricity sales data provided by SDG&E to EPIC, August 16, 2018.

16 Loss factor is from CEC Energy Demand 2019 Forecast. For each forecast cycle, utilities provide the estimates, which remain relatively stable. Personal communication with CEC staff. March 23, 2020.

17 EPIC: Estimating annual average greenhouse gas emission factors for the electric sector: a method for inventories (2016), accessed May 7, 2020.

18 D.14-12-037, December 18, 2014, in Rulemaking 11-03-012 (filed March 24, 2011). The recommended emission factor is 0.379 MT CO_2e/MWh (836 lbs CO_2e/MWh).

Information Administration (EIA) Form 923 data, and the 2016 SDG&E Power Source Disclosure Program to identify the self-serve electric generation plants.

With the adjustments, the key inputs and results are shown in Table X.11.

Table X.11: Key Inputs and 2016 Greenhouse Gas Emissions from Electricity

Key Inputs and 2016 Greenhouse Gas Emissions from Electricity	
Electricity Sales – Bundled (MWh)	14,482,332
Electricity Sales – Direct Access (MWh)	3,360,561
Transmission and Distribution Loss Factor	1.082
SDG&E Electricity Emission Factor (lbs CO_2e/MWh)	527
Direct Access Electricity Emission Factor (lbs CO_2e/MWh)	836
GHG Emissions (MT CO_2e)	5,121,950
GHG Emissions Associated with Electricity for Water Treatment – Excluded (MT CO_2e)	−58,925
GHG Emissions Associated with Natural Gas Used at On-Site Self-Serve Electric Generation – Added (MT CO_2e)	204,014
GHG Emissions (MT CO_2e)	5,267,039
GHG Emissions (MMT CO_2e)	**5.3**

Source: CEC 2020; SDG&E 2018; EPIC, USD 2020.

Difference from Previous 2012 Inventory

Methods to estimate emissions from electricity are the same in both the 2012 and 2016 GHG inventories. However, source data have been updated and refined. For example, the DA emission factor was not available for the 2012 inventory but was available for the 2016 inventory.

Method Used to Develop Emissions Projections

To project emissions for the electricity category, EPIC estimated the impact of federal and state policies and regulations on separately reducing electricity use and reducing the electricity emission factor (by increasing renewable or zero-carbon electricity).

Senate Bill 100 (De León, 2018) (Chapter 312, Statutes of 2018) (SB 100), the 100 Percent Clean Energy Act of 2018, increases California's Renewable Portfolio Standard (RPS) to 60% by 2030.[19] The legislation also provides goals for the years leading up to 2030 and establishes a state policy requiring eligible renewable resources and zero-carbon resources to supply 100% of all retail electricity sales by 2045. All retail electricity providers

[19] California Senate Bill 100 (De León, 2018) (Chapter 312, Statutes of 2018).

must meet these RPS requirements, including utilities (e.g., SDG&E), ESPs for DA customers, and other local renewable programs (e.g., Community Choice Energy [CCE] programs). EPIC assumed that all retail electricity providers will meet the 2030 and 2045 SB 100 targets.

In addition, San Diego Community Power (SDCP), a CCE program formed by the cities of Chula Vista, Encinitas, Imperial Beach, La Mesa, and San Diego, started delivering power in March 2021. SDCP plans to start with 55% GHG-free electricity in 2021 and to supply 100% renewable electricity by 2030 or 2035.[20] Because SDCP will be operational by time the 2021 Regional Plan is adopted, the impact of SDCP delivering GHG-free electricity above the 2030 RPS target is included in the emissions projection. Another CCE program, Clean Energy Alliance (CEA), formed by the cities of Carlsbad, Del Mar, and Solana Beach, started delivering power in May 2021. Because the planned renewable content in CEA's electricity supply is consistent with the RPS target, 60% renewable by 2030, the impact of the CEA is not shown separately.[21] The projected renewable or GHG-free content and electricity emission factors for each supplier are shown in Table X.12.

Table X.12: Projected Renewable or Greenhouse Gas–Free Content and Emission Factors of Retail Electricity Providers

Projected Renewable or Greenhouse Gas–Free Content and Emission Factors of Retail Electricity Providers

Retail Electricity Provider	2025	2030	2035	2045	2050
Projected Renewable or GHG-Free Content (%)*					
San Diego Community Power	67%	100%	100%	100%	100%
SDG&E Bundled and Clean Energy Alliance	47%	60%	73%	100%	100%
ESPs for Direct Access	47%	60%	73%	100%	100%
Projected Electricity Emission Factor (lbs CO_2e/MWh)**					
San Diego Community Power	308	—	—	—	—
SDG&E Bundled and Clean Energy Alliance	493	370	249	—	—
ESPs for Direct Access	493	370	249	—	—

* Based on SB 100 RPS targets and CCE programs' implementation plans.

** Calculated based on 2016 SDG&E bundled electricity emission factor of 527 lbs CO_2e/MWh and 43% renewable provided in its 2018 Power Source Disclosure.

Source: EPIC, USD 2020.

[20] SDCP: Community Choice Aggregation Implementation Plan and Statement of Intent (2019), accessed August 4, 2020. SDCP: Board of Directors Meeting, May 28, 2020, SDCP Renewable and GHG-Free Targets, accessed August 4, 2020.

[21] CEA: Community Choice Aggregation Implementation Plan and Statement of Intent (2019), accessed December 22, 2020.

The latest CEC California Energy Demand 2020–2030 Revised Forecast projects electricity sales in the SDG&E planning area (service area) through 2030. The electricity sales account for the impact of behind-the-meter photovoltaic (PV) and non-PV self-generation, behind-the-meter storage, current electricity rate structure, and appliance and building energy efficiency standards up to 2019.[22] EPIC applied the rate of increase from CEC's Demand Forecast electricity sales projection for the SDG&E planning area to the 2016 San Diego region's electricity sales. As no forecast is available for after 2030, EPIC used the 2029–2030 annual electricity sales increase, 0.7%, as the post-2030 annual increase. Assuming existing DA customers remain and there are no additional new retail electricity suppliers in the San Diego region, the projected electricity sales by supplier are shown in Table X.13.

Table X.13: Projected Electricity Sales of Retail Electricity Providers

Projected Electricity Sales of Retail Electricity Providers					
Retail Electricity Provider	**2025**	**2030**	**2035**	**2045**	**2050**
Projected Electricity Sales (GWh)					
San Diego Community Power*	7,408	7,189	7,459	8,031	8,333
SDG&E Bundled and Clean Energy Alliance	5,775	6,403	6,137	5,573	5,275
ESPs for Direct Access	3,059	3,154	3,155	3,157	3,158

* Estimated based on the projected demand through 2030 in SDCP Implementation Plan and SDG&E Planning Area electricity sales in CEC 2020–2030 energy demand forecast, 2021 version.

Source: EPIC, USD 2020.

With the projected electricity sales and emission factor of each supplier, assuming 2016 self-serve co-generation plants will still be operational at existing levels in the forecast years, the projected emissions are shown in Table X.14.

[22] CEC: Final 2020 Integrated Energy Policy Report Update Volume III: California Energy Demand Forecast Update (March 2021).

Table X.14: Projected Greenhouse Gas Emissions from Electricity

Projected Greenhouse Gas Emissions from Electricity					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
GHG Emissions from Electricity Sales (MT CO_2e)*	3,256,139	1,733,379	1,137,543	—	—
GHG Emissions from Water Treatment Excluded (MT CO_2e)	68,048	53,095	37,058	—	—
GHG Emissions from On-Site Self-Serve Electricity Generation Included (MT CO_2e)	204,104	204,104	204,104	204,104	204,104
Adjusted GHG Emissions (MT CO_2e)	3,392,104	1,884,298	1,304,499	204,104	204,014
GHG Emissions (MMT CO_2e)	**3.4**	**1.9**	**1.3**	**0.2**	**0.2**

* Electricity sales from SDCP, SDG&E, CEA, and ESPs for DA.

Source: EPIC, USD 2020.

Natural Gas

The combustion of natural gas for building end use accounts for 12% of total emissions in the 2016 inventory and 20% in the 2050 projection. This category calculates emissions from building end use natural gas for purposes other than electric generation, not for utility-level electric generation (UEG) and not for on-site self-serve electric generation, as they are accounted for under the electricity category. However, emissions associated with natural gas use for heat output from any of the co-generation plants are captured in this category.

Method Used to Estimate 2016 Emissions

To estimate GHG emissions from metered natural gas end use, EPIC multiplied the metered natural gas sales by a constant natural gas emission factor.

SDG&E provided the 2016 San Diego regional natural gas sales by customer class. The San Diego regional natural gas sales are sales to all local jurisdictions, including military bases and tribal reservations. The natural gas use for UEG purposes, either at co-generation or electric generation plants, is excluded.[23] However, certain co-generation plants may have dual purposes that generate electricity use for both on-site use and sales to the utility. EPIC used the natural gas emission factor, 0.00545 MT CO_2e per therm, based on CARB's statewide inventory data.[24]

23 Natural gas sales data provided by SDG&E, August 16, 2018.

24 CARB: Documentation of California's Greenhouse Gas Inventory (11th Edition), accessed March 23, 2020. The natural gas emission factor is also used in CARB Mandatory GHG Reporting and is the same under each customer class (e.g., residential, commercial).

Three adjustments are made to the emissions estimate based on natural gas sales:

- Emissions associated with natural gas used at on-site self-serve electric generation, mostly co-generation plants, were removed from this category and allocated to the electricity category. EPIC used the CEC QFER Power Plant Owner Reporting database, EIA Form 923, and the 2016 SDG&E Power Source Disclosure Program to identify the self-serve electric generation plants.
- Emissions associated with natural gas used for utility electric sales at dual-purpose (both on-site use and utility sales) co-generation plants were removed from this category because they are already accounted for in the electricity emission factor calculation. The method to identify the plants is the same as above.
- Emissions associated with heat output from utility-level co-generation plants were estimated separately and added to this category. This natural gas use is not captured in the SDG&E natural gas sales. EPIC assumed that excess heat output was sold by the plants for other use (e.g., to another industrial customer nearby). The method to identify the plants is the same as above.

With these adjustments, the key inputs and results are shown in Table X.15.

Table X.15: Key Inputs and 2016 Greenhouse Gas Emissions from Natural Gas

Key Inputs and 2016 Greenhouse Gas Emissions from Natural Gas	
Natural Gas Sales (therms)	585,460,937
Natural Gas Emission Factor (MT CO_2e/therm)	0.00545
GHG Emissions (MT CO_2e)	3,192,578
GHG Emissions Associated with Heat Output from Utility-Level Co-Generation Plants – Included (MT CO_2e) (1)	118,239
GHG Emissions from Natural Gas used to Generate Electricity for Sales to Utility – Excluded (MT CO_2e)* (2)	−3,593
GHG Emissions from Natural Gas Used at On-Site Self-Serve Electric Generation – Excluded (MT CO_2e) (3)	−204,014
Total Adjustment (MT CO_2e) (1+2+3)	−89,369
GHG Emissions (MT CO_2e)	3,103,209
GHG Emissions (MMT CO_2e)	**3.1**

* Does not include power plants generating electricity for utility sales only.

Source: CARB 2019; SDG&E 2018; EPIC, USD 2020.

Difference from Previous 2012 Inventory

Methods to estimate emissions from natural gas are the same in both the 2012 and 2016 inventories. However, the source data (e.g., the data associated with co-generation plants in the San Diego region) have been updated and refined.

Method Used to Develop Emissions Projections

To project emissions for the natural gas category, EPIC estimated the impact of federal and state polices and regulations on reducing natural gas use. The natural gas emission factor, 0.00545 MT CO_2e per therm, is a constant.

The 2020 version of the CEC California Energy Demand 2020–2030 Forecast projects natural gas sales in the SDG&E planning area through 2030.[25] The natural gas sales already account for the impact of the current natural gas rate structure, as well as appliance and building energy efficiency standards up to 2019. Unlike SDG&E's electricity service area, SDG&E's natural gas service area matches the boundaries of the San Diego region. EPIC applied the rate of increase from the CEC Demand Forecast for the SDG&E planning area to 2016 natural gas sales for the San Diego region. Since no forecast is available after 2030, EPIC used the 2029–2030 annual natural gas sales increase, 0.2%, as a post-2030 annual increase. Assuming the 2016 co-generation plants adjustment does not change, the projected emissions are shown in Table X.16.

25 The CEC Energy Demand Forecast has a one-year cycle for the electricity demand forecast, but a two-year cycle for the natural gas demand forecast.

Table X.16: Projected Greenhouse Gas Emissions from Natural Gas

Projected GHG Emissions from Natural Gas					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Projected Natural Gas Sales (therms)*	628,689,290	640,276,291	647,766,840	663,011,857	670,768,387
Natural Gas Emission Factor (MT CO_2e/therm)	0.00545	0.00545	0.00545	0.00545	0.00545
GHG Emissions from Natural Gas Sales (MT CO_2e)	3,428,892	3,492,088	3,532,942	3,616,089	3,658,393
Total Adjustment for Co-Generation Plants (MT CO_2e)**	−89,369	−89,369	−89,369	−89,369	−89,369
GHG Emissions (MT CO_2e)	3,339,523	3,402,719	3,443,573	3,526,720	3,569,024
GHG Emissions (MMT CO_2e)	3.3	3.4	3.4	3.5	3.6

* Estimated based on CEC 2020–2030 energy demand forecast, 2020 version.

** Calculated in Table X.15.

Source: EPIC, USD 2020.

Industrial

Emissions from GHGs with high GWPs used in industrial processes and the processing of materials to manufacture items (e.g., mineral aggregate products, chemicals, metals, refrigerants, electronics, and other consumer goods) account for 8% of total emissions in the 2016 inventory and 14% in the 2050 projection. GHGs with high GWPs are used in air conditioning units and refrigeration, as well as in the manufacturing of electronics, fire protection equipment, insulation, and aerosols. This category focuses on industrial processes that directly release CO_2 and other GHGs with high GWPs (i.e., SF_6, C_2F_6, C_3F_8, CF_4, C_4F_8, HFC-23, NF_3, HFC-125, HFC-134a, HFC-143a, HFC-236fa, HFC-32) by processes other than fuel consumption.

Method Used to Estimate 2016 Emissions

Similar to the method used in the other fuels category, EPIC scaled down the industrial emissions in the CARB statewide GHG inventory to the San Diego region based on the San Diego region to state ratio relevant to each economic sector.[26]

The following are the IPCC category numbers, subcategory numbers, headings, codes, and fuel types used within each type of activity in the statewide inventory. Only those categories, subcategories, activities, and fuel types causing emissions in the San Diego region are shown:

- 2D1: Industrial Lubricant Use
  - Not Specified Industrial > Fuel Consumption – Lubricants > CO_2
  - Not Specified Transportation > Fuel Consumption – Lubricants > CO_2
- 2D3: Industrial Solvent Use
  - Solvents & Chemicals: Evaporative Losses: Fugitives > Fugitive Emissions > CO_2
- 2E: Electronic Industry
  - Manufacturing: Electric & Electronic Equip.: Semiconductors & Related Products > Semiconductor Manufacture > C_2F_6
  - Manufacturing: Electric & Electronic Equip.: Semiconductors & Related Products > Semiconductor Manufacture > C_3F_8
  - Manufacturing: Electric & Electronic Equip.: Semiconductors & Related Products > Semiconductor Manufacture > C_4F_8
  - Manufacturing: Electric & Electronic Equip.: Semiconductors & Related Products > Semiconductor Manufacture > CF_4
  - Manufacturing: Electric & Electronic Equip.: Semiconductors & Related Products > Semiconductor Manufacture > HFC-23
  - Manufacturing: Electric & Electronic Equip.: Semiconductors & Related Products > Semiconductor Manufacture > NF_3
  - Manufacturing: Electric & Electronic Equip.: Semiconductors & Related Products > Semiconductor Manufacture > SF_6
- 2F: Product Uses as – Not Specified Commercial
  - Use of Substitutes for Ozone-Depleting Substances > CF_4
  - Use of Substitutes for Ozone-Depleting Substances > HFC-125
  - Use of Substitutes for Ozone-Depleting Substances > HFC-134a
  - Use of Substitutes for Ozone-Depleting Substances > HFC-143a

[26] CARB: CARB Greenhouse Gas Emission Inventory – Query Tool for years 2000 to 2017 (12th edition), accessed on June 5, 2020.

- Use of Substitutes for Ozone-Depleting Substances > HFC-236fa
    - Use of Substitutes for Ozone-Depleting Substances > HFC-32
    - Use of Substitutes for Ozone-Depleting Substances > Other Ozone-Depleting Substances Substitutes
- 2G1b: Other Industrial Product – Electrical
    - Imported Electricity: Transmission and Distribution > Electricity Transmitted > SF_6
    - In State Generation: Transmission and Distribution > Electricity Transmitted > SF_6
- 2G4: Other Industrial Product – CO_2, Limestone
    - Not Specified Industrial > CO_2 Consumption > CO_2
    - Not Specified Industrial > Limestone and Dolomite Consumption > CO_2
    - Not Specified Industrial > Soda Ash Consumption > CO_2

EPIC used different ratios to scale down the activities above to the San Diego region. Table X.17 shows the ratios used and their values in 2016.

Table X.17: Key Inputs and 2016 Greenhouse Gas Emissions from Industrial

Key Inputs and 2016 Greenhouse Gas Emissions from Industrial				
Economic Sector/Industry	**Basis for Ratio Value**	**California (MMT CO_2e)**	**Ratio Value**	**San Diego Region (MMT CO_2e)**
Industrial Lubricant and Limestone Use	San Diego manufacturing sector employees/ California manufacturing sector employees	1.93	9%	0.17
Industrial Lubricant Use – Not Specified Transportation (Lubricant, ODS)	San Diego VMT/California statewide VMT	5.55	9%	0.51
Industrial Solvent Use – Solvents and Chemicals	San Diego manufacturing sector employees/ California manufacturing sector employees	0.79	9%	0.07
Electronic Industry – Semiconductor Manufacture	San Diego semiconductor manufacturing sector employees/California semiconductor manufacturing sector employees	0.16	7%	0.01
Not Specified Residential (ODS)	San Diego total residential units/California total residential units	3.17	9%	0.27
Not Specified Commercial (ODS)	San Diego total employees/California total employees	11.9	9%	1.01
Imported Electricity – Transmission and Distribution	San Diego purchased electricity/California purchased electricity	0.03	11%	0.004
In State Generation – Transmission and Distribution	San Diego in-county electricity generated/ California in-state electricity generated	0.07	3%	0.002
Total GHG Emissions (MMT CO_2e)		**24**	**N/A**	**2.1**

ODS – Emissions from use of substitutes for Ozone-Depleting Substances.

Source: 2016 County Business Patterns; SANDAG ABM14.2.2 VMT; EMFAC2017 statewide on-road emission inventory; SANDAG demographic data; EPIC, USD 2021.

Emissions from the following categories were included in CARB's statewide inventory but not in the 2016 regional inventory because Economic Census data indicated no economic activity in the San Diego region.[27] The categories are:

- 2A1: Manufacturing: Stone, Clay, Glass, and Cement: Cement > Clinker Production > CO_2
- 2A2: Manufacturing: Stone, Clay, Glass, and Cement: Lime > Lime Production > CO_2
- 2B2: Manufacturing: Chemical and Allied Products: Nitric Acid > Nitric Acid Production > N_2O
- 2H3: Petroleum Refining: Transformation > Fuel Consumption > CO_2

Difference from Previous 2012 Inventory

Methods to estimate emissions from the industrial sector are the same in both the 2012 and 2016 inventories.

Similar to the other fuels category, there are no empirical data for industrial activities in the San Diego region. For the 2016 inventory, EPIC used the same methodology as the 2012 inventory. However, refinements were made on the downscaling ratios. For industrial (not specified) lubricant use, the 2012 inventory used the VMT ratio. In the 2012 inventory, the emissions, due to use of substitutes for ODS, were a single category and were scaled down based on the population ratio. For the 2016 inventory, EPIC used CARB's categories and categorized these emissions into not-specified transportation, not-specified commercial, and not-specified residential sectors. The ratios to scale down these emissions were discussed in the previous section. For the emissions due to soda ash and limestone consumption, which is a not-specified industrial activity, EPIC used the ratio of the manufacturing sector employees instead of the ratio of population used in the 2012 inventory.

Method Used to Develop Emissions Projections

EPIC projected emissions for the industrial sector are based on the San Diego regional population, housing, jobs, and VMT projections. Each specific industry is projected separately based on the type of activity as shown in Table X.17. For example, the emissions from transportation lubricants use were projected based on San Diego regional VMT forecast, and the emissions from solvents and chemicals were projected based on the San Diego regional manufacturing jobs forecast. The projected emissions are shown in Table X.18.

[27] Confirmed by San Diego Economic Development Corporation research team, personal communication.

Table X.18: Projected Greenhouse Gas Emissions from Industrial

Projected Greenhouse Gas Emissions from Industrial					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Manufacturing Sector Jobs Increase Compared with 2016 (%)	15%	21%	26%	34%	37%
Population Increase Compared with 2016 (%)	6%	8%	10%	13%	14%
VMT Increase Compared with 2016 (%)	1%	2%	2%	4%	5%
Housing Increase Compared with 2016 (%)	9%	14%	19%	23%	24%
Jobs Increase Compared with 2016 (%)	12%	12%	17%	24%	27%
Total GHG Emissions (MMT CO_2e)	**2.2**	**2.3**	**2.3**	**2.5**	**2.5**

Source: SANDAG 2021; EPIC, USD 2021.

On-Road Transportation – Heavy-Duty Trucks and Vehicles

The on-road transportation heavy-duty trucks and vehicles category accounts for 7% of total GHG emissions in the 2016 inventory and 10% in the 2050 projection. Vehicle classes included in this category are taken from EMFAC2017.[28]

Method Used to Estimate 2016 Emissions

EPIC used the same method to estimate emissions from this category and the on-road transportation passenger cars and light-duty vehicles category, with an EMFAC2017 model run of VMT from SANDAG ABM14.2.2 and tons of CO_2 per weekday to MT CO_2e per year conversion. The key inputs and results are shown in Table X.19.

[28] Vehicle classes are all except LDA, LDT1, LDT2, and MDV as shown in EMFAC2017 Technical Documentation, Table 6.1-1.

Table X.19: Key Inputs and 2016 Greenhouse Gas Emissions from On-Road Transportation – Heavy-Duty Trucks and Vehicles

Key Inputs and 2016 Greenhouse Gas Emissions from On-Road Transportation – Heavy-Duty Trucks and Vehicles	
VMT (miles per weekday)*	4,834,783
CO_2 Emissions (tons per weekday)**	5,866
Conversion Factor (tons CO_2 per weekday to MT CO_2e per year)	300
GHG Emissions (MT CO_2e)	1,761,445
GHG Emissions (MMT CO_2e)	**1.8**

* SANDAG ABM14.2.2 VMT.

** EMFAC2017 model run with custom VMT inputs from SANDAG.

Heavy-duty trucks and vehicles are EMFAC2017 vehicle categories except LDA, LDT1, LDT2, and MDV. Conversion factors are different for each vehicle class.

Source: CARB 2016, 2017; SANDAG 2021; EPIC, USD 2021.

Difference from Previous 2012 Inventory

Methods to estimate emissions from heavy-duty trucks and vehicles are the same in both 2012 and 2016 inventories. However, the previous versions of the SANDAG ABM and EMFAC2014 were used to calculate these emissions in the 2012 GHG inventory. The SANDAG ABM has undergone changes, and EMFAC2017 includes new regulations that were not reflected in EMFAC2014.

Method Used to Develop Emissions Projections

The method used to develop the GHG projections for heavy-duty trucks and vehicles is the same as the method used to project emissions from passenger cars and light-duty vehicles. The new and updated regulations for heavy-duty trucks and vehicles since the release of EMFAC2014 are:

- Senate Bill 1 (Beall, 2017) (the Road Repair and Accountability Act of 2017) and the CARB Tractor Trailer GHG Regulation require medium-duty or heavy-duty vehicles to verify compliance with CARB's Truck and Bus Regulation. EMFAC2017 assumes full compliance by 2023. CARB's Tractor-Trailer GHG Regulation includes aerodynamic and tire improvement requirements to reduce GHG emissions from heavy-duty trucks.
- U.S. EPA's Phase 2 GHG Regulation for heavy-duty vehicles (heavy-duty trucks, tractors, and buses) built upon the Phase 1 standards with new requirements beginning with model year 2018 for trailers and model year 2021 for engines and vehicles, with phase-in through model year 2027.[29]

[29] CARB: EMFAC2017 Volume III – Technical Documentation (2018), accessed April 30, 2020.

Using the same conversion method from tons of CO_2 per weekday to MT CO_2e per year discussed in the inventory method section, the key inputs and results are shown in Table X.20.

Table X.20: Key Inputs and Projected Greenhouse Gas Emissions from On-Road Transportation – Heavy-Duty Trucks and Vehicles

Key Inputs and Projected Greenhouse Gas Emissions from On-Road Transportation – Heavy-Duty Trucks and Vehicles					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
VMT (miles per weekday)*	5,245,381	5,640,701	5,912,387	6,377,437	6,573,289
CO_2 Emissions (tons per weekday)**	5,620	5,618	5,548	5,649	5,714
Conversion Factor (MT CO_2e per year/ tons per weekday)	299	299	299	299	299
GHG Emissions (MT CO_2e)	1,682,590	1,681,227	1,659,696	1,690,091	1,709,802
GHG Emissions (MMT CO_2e)	1.7	1.7	1.7	1.7	1.7

* SANDAG ABM14.2.2 VMT.

** EMFAC2017 model run with custom VMT inputs from SANDAG ABM14.2.2.

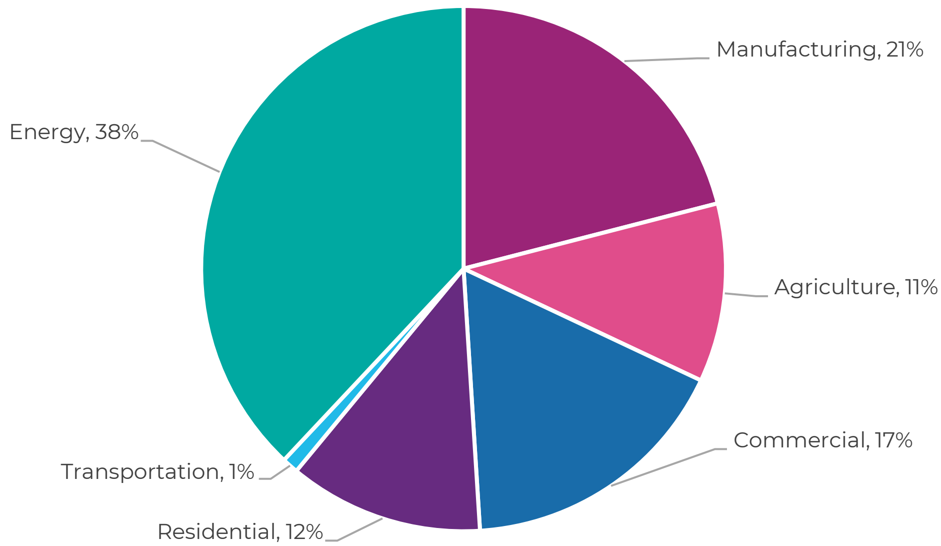
Source: CARB 2016, 2017; SANDAG 2021; EPIC, USD 2021.

Other Fuels

The Other Fuels category accounts for 4% of total emissions in the 2016 inventory and 9% in the 2050 projection. These fuels include distillate (other than in power production), kerosene, gasoline (other than in transportation), liquefied petroleum gas (LPG), residual fuel oil (other than in power production), and wood (wet).

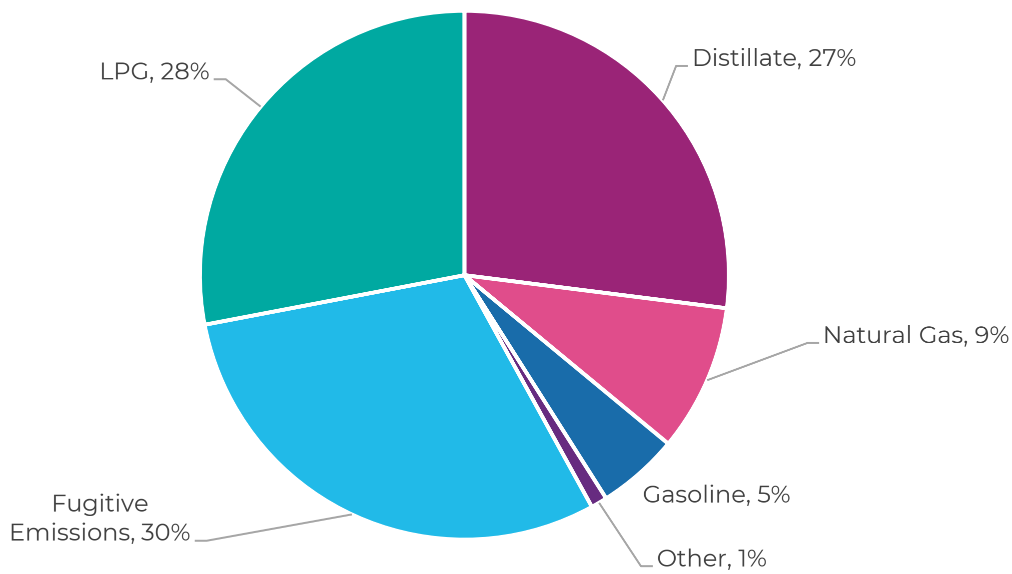
Emissions from this category are divided into the following economic sectors, according to the CARB statewide GHG inventory: agriculture, commercial, residential, transport, energy, and manufacturing. The relative distribution of emissions by economic sector is provided in Figure X.1 and by fuel type in Figure X.2.

Figure X.1: Relative Distribution of 2016 Greenhouse Gas Emissions from Other Fuels by Economic Sectors



Source: EPIC, USD 2020.

Figure X.2: Relative Distribution of 2016 Greenhouse Gas Emissions from Other Fuels by Fuel Type



Source: EPIC, USD 2020.

Method Used to Estimate 2016 Emissions

The GHG emissions from the CARB statewide inventory were the basis of the regional estimates.[30] EPIC scaled down the statewide emissions by economic sector to the San Diego region based on whether a particular category had any economic activity in the San Diego region using relevant economic, population, employment, or transportation data. Therefore, not all of CARB's statewide emissions from these economic sectors are included in the 2016 regional inventory.

CARB uses the IPCC category and subcategory names and codes, as specified in the IPCC 2006 Guidelines for GHG Inventories, to be consistent with the EPA national inventory. Below are only those IPCC categories, subcategories, activities, and fuel types with GHG emissions in the San Diego region, based on economic activity data in the San Diego region.

CARB agriculture sector: EPIC scaled down the emissions from the following categories to San Diego region using the 2016 ratio of the revenue generated by agricultural activities in the San Diego region to the statewide agricultural revenue.[31]

- 1A4c: Agriculture/Forestry/Fishing/Fish Farms > Agriculture Energy Use
  - Distillate > CH_4, CO_2, N_2O
  - Kerosene > CH_4, CO_2, N_2O
  - Gasoline > CH_4, CO_2, N_2O
  - Ethanol > CH_4, CO_2, N_2O

CARB commercial sector: EPIC scaled down the emissions from the following categories to the San Diego region using the 2016 ratio of the number of employees in the San Diego region's manufacturing sector to the statewide manufacturing sector.[32]

- 1A4a: Commercial/Institutional > Not Specified Commercial
  - Distillate > CH_4, CO_2, N_2O
  - Kerosene > CH_4, CO_2, N_2O
  - Gasoline > CH_4, CO_2, N_2O
  - LPG > CH_4, CO_2, N_2O
  - Residual Fuel Oil > CH_4, CO_2, N_2O
  - Wood (wet) > CH_4, N_2O

30 CARB Greenhouse Gas Emission Inventory – Query Tool for years 2000 to 2017 (12th edition), accessed on May 25, 2020.

31 California Department of Food & Agriculture: California Agricultural Statistics Review, 2016–2017. accessed May 28, 2020.

32 2016 County Business Patterns, accessed on May 30, 2020. The 2012 North American Industry Classification System (NAICS) Code for manufacturing Sector is 31-33.

CARB residential sector: EPIC scaled down the emissions from the following categories to the San Diego region using the 2016 ratio of the San Diego regional population to the statewide population.[33]

- 1A4b: Residential > Household Use
  - Distillate > CH_4, CO_2, N_2O
  - Kerosene > CH_4, CO_2, N_2O
  - LPG > CH_4, CO_2, N_2O
  - Wood (wet) > CH_4, N_2O

CARB transportation sector: This category included the emissions from LPG fuel combustion. EPIC scaled down the emissions from the following categories to the San Diego region using the 2016 ratio of San Diego regional VMT to statewide VMT.[34]

- 1A3: Transport > Not Specified Transportation
  - LPG > CH_4, CO_2, N_2O
  - Residual Fuel Oil > CH_4, CO_2, N_2O

CARB energy sector: This category included the emissions from the transmission and distribution of electricity (e.g., fugitive and fuel combustion emissions from natural gas pipelines used for electric generation, non-natural gas pipelines and natural gas storage). EPIC scaled down the emissions from the following categories to the San Diego region using the 2016 ratio of total establishments for transmission and distribution activities in the San Diego region to the statewide establishments for the same activities.[35]

- 1B2: Oil and Natural Gas
  - Not Specified Industrial > Fugitives > Fugitive Emissions > CH_4
  - Pipelines > Natural Gas > Fugitives > Fugitive Emissions > CH_4, CO_2
- 1A1: Energy Industries > Pipelines
  - Natural Gas Pipelines > Natural Gas > CH_4, CO_2, N_2O
  - Non-Natural Gas Pipelines > Natural Gas > CH_4, CO_2, N_2O

---

[33] San Diego demographic data are shown in Table X.2. Statewide population projections are from California Department of Finance, accessed on May 30, 2020.

[34] San Diego regional 2016 VMT are provided in Table X.4 and Table X.19. California statewide VMT is from EMFAC2017, accessed on June 1, 2020.

[35] 2016 County Business Patterns, accessed on May 30, 2020. The 2012 NAICS Code for Electric Power Generation, Transmission and Distribution is 2211.

CARB manufacturing sector: EPIC scaled down the emissions from the following categories to the San Diego region using the 2016 ratio of the number of employees in the San Diego region's manufacturing sector and the statewide manufacturing sector.[36]

- 1A2f: Manufacturing Industries and Construction > Non-Metallic Minerals > Stone, Clay, Glass, and Cement > Cement
  - Distillate > CH_4, CO_2, N_2O
  - LPG > CH_4, CO_2, N_2O
  - MSW > CH_4, CO_2, N_2O
  - Petroleum Coke > CH_4, CO_2, N_2O
  - Residual Fuel Oil > CH_4, CO_2, N_2O
  - Tires > CH_4, CO_2, N_2O
- 1A2k: Manufacturing Industries and Construction > Construction
  - Gasoline > CH_4, CO_2, N_2O
- 1A2m: Manufacturing Industries and Construction > Non-Specified Industry
  - Distillate > CH_4, CO_2, N_2O
  - Gasoline > CH_4, CO_2, N_2O
  - Kerosene > CH_4, CO_2, N_2O
  - LPG > CH_4, CO_2, N_2O
  - Petroleum Coke > CH_4, CO_2, N_2O
  - Residual Fuel Oil > CH_4, CO_2, N_2O
- 1B2: Oil and Natural Gas > Manufacturing
  - Chemicals and Allied Products > Fugitives > Fugitive Emissions > CH_4
  - Construction > Fugitives > Fugitive Emissions > CH_4
  - Electric and Electronic Equipment > Fugitives > Fugitive Emissions > CH_4
  - Food Products > Fugitives > Fugitive Emissions > CH_4
  - Fugitives > Fugitive Emissions > CH_4
  - Plastic and Rubber > Fugitives > Fugitive Emissions > CH_4
  - Primary Metals > Fugitives > Fugitive Emissions > CH_4
  - Pulp and Paper > Fugitives > Fugitive Emissions > CH_4
  - Storage Tanks > Fugitives > Fugitive Emissions > CH_4

[36] 2016 County Business Patterns, accessed on May 30, 2020. The 2012 NAICS code for manufacturing sector is 31-33.

Several categories were included in CARB's statewide inventory, but not in this 2016 regional inventory, because 2016 business patterns in data for the San Diego region indicated no economic activities under these categories. The categories are:

- 1A1b: Petroleum Refining
    - Associated Gas > CH_4, CO_2, N_2O
    - Catalyst Coke> CH_4, CO_2, N_2O
    - Distillate> CH_4, CO_2, N_2O
    - LPG > CH_4, CO_2, N_2O
    - Petroleum Coke > CH_4, CO_2, N_2O
    - Refinery Gas > CH_4, CO_2, N_2O
    - Residual Fuel Oil > CH_4, CO_2, N_2O
- 1A1c: Manufacture of Solid Fuels and Other Energy Industries
    - Associated Gas > CH_4, CO_2, N_2O
    - Crude Oil > CH_4, CO_2, N_2O
    - Distillate > CH_4, CO_2, N_2O
    - Residual Fuel Oil > CH_4, CO_2, N_2O
- 1B2: Oil and Natural Gas > Manufacturing: Stone, Clay, Glass, and Cement: Fugitives > Fugitive Emissions > CH_4
- 1B2a: Oil > Petroleum Refining: Process Losses: Fugitives > Fugitive Emissions > CH_4
- 1B3: Other Emissions from Energy Production > In State Generation: Merchant Owned > Geothermal Power – Geothermal > CO_2
- 1B3: Other Emissions from Energy Production > In State Generation: Utility Owned > Geothermal Power > CO_2

The key inputs and results are shown in Table X.21.

Table X.21: Key Inputs and 2016 Greenhouse Gas Emissions from Other Fuels

Key Inputs and 2016 Greenhouse Gas Emissions from Other Fuels	
Economic Sectors Associated with Other Fuels*	**2016 Emissions (MMT CO_2e)**
Agriculture	0.12
Commercial	0.20
Residential	0.13
Transportation	0.01
Energy	0.44
Manufacturing	0.24
Total GHG Emissions	**1.1**

* Economic sectors used in CARB statewide GHG inventory.

Source: California Agricultural Statistics Review 2016–2017; 2016 County Business Patterns; SANDAG ABM14.2.2 VMT; EMFAC2017 statewide on-road emission inventory; EPIC, USD 2020.

Difference from Previous 2012 Inventory

Methods to estimate emissions from other fuels are the same in both the 2012 and 2016 inventories. However, estimates were refined with different ratios to scale down the statewide emissions. For the energy sector, the ratio of establishments for transmission and distribution activities was used for the 2016 inventory instead of the ratio of energy consumption in the 2012 inventory, because emissions under this sector are due to the transmission and distribution pipelines. For the manufacturing sector, the ratio of the number of employees in the manufacturing sector was used for the 2016 inventory instead of the ratio of the total employees in all sectors, so the emissions are specific to the manufacturing sector.

Method Used to Develop Emissions Projections

Except for the agriculture sector, EPIC projected emissions for the other fuels sector based on the San Diego regional population, jobs, and VMT projections. The projected emissions associated with the manufacturing, energy, and commercial sectors were based on the jobs forecast. The projected emissions associated with the residential sector were based on the population forecast. The projected emissions associated with the transportation sector were based on the VMT forecast.

For the agriculture sector, EPIC used Microsoft Excel's GROWTH function to project San Diego regional and statewide agriculture revenue. The GROWTH function predicts the growth with existing data. The projected emissions for the agriculture sector were based on the annual growth rate of ratio of San Diego region to California agriculture revenue. The projected emissions are shown in Table X.22.

Table X.22: Projected Greenhouse Gas Emissions from Other Fuels

Projected Greenhouse Gas Emissions from Other Fuels					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Total Agricultural GHG Emissions (MMT CO_2e)	0.12	0.10	0.08	0.08	0.06
Total Commercial GHG Emissions (MMT CO_2e)	0.20	0.23	0.24	0.25	0.26
Total Residential GHG Emissions (MMT CO_2e)	0.13	0.14	0.14	0.15	0.15
Total Transportation GHG Emissions (MMT CO_2e)	0.01	0.01	0.01	0.01	0.01
Total Energy (Electricity Transmission and Distribution) GHG Emissions (MMT CO_2e)	0.44	0.61	0.64	0.67	0.71
Total Manufacturing GHG Emissions (MMT CO_2e)	0.24	0.28	0.29	0.31	0.33
Total GHG Emissions (MMT CO_2e)	**1.4**	**1.4**	**1.5**	**1.5**	**1.5**

Source: EPIC, USD 2020.

Off-Road Transportation

The off-road transportation category includes the following subcategories by equipment type: construction and mining equipment, cargo handling equipment, industrial equipment, airport ground support, pleasure craft, recreational equipment, lawn and garden equipment, agricultural equipment, transport refrigeration units, military tactical support equipment, and other portable equipment. The GHG emissions from off-road equipment fuel combustion account for 2% of total emissions in the 2016 inventory and 5% in the 2050 projection.

Method Used to Estimate 2016 Emissions

CARB released the OFFROAD ORION model in 2017 and the SORE model in 2020.[37] The ORION 2017 model generates off-road equipment emission data by county, vehicle category, vehicle type, horsepower (HP), and fuel type. SORE 2020 is a standalone Microsoft Access model that generates emission data for off-road vehicles with engines less than or equal to 25 HP. EPIC used ORION 2017 to generate 2016 regional off-road emissions for HP greater than or equal to 25. For the vehicles with HP equal to 25, data may overlap with SORE 2020 results. EPIC used SORE 2020 results for the overlapping vehicles because SORE 2020 is the latest and most recently updated model. Pleasure crafts and recreational vehicles are subcategories in ORION 2017; however, no San Diego regional data were available. EPIC used CARB's pleasure craft model, PC2014, and recreational vehicle model, RV 2018, to generate the emission data for the respective subcategories.[38] Like SORE 2020, both these models are standalone Microsoft Access models.

---

37 CARB: ORION 2017 and SORE 2020 Small Off Road Engine model.

38 CARB: PC2014 Pleasure Craft model and RV 2018 Recreational Vehicle model.

Table X.23 shows the different databases used to generate the emissions for the different vehicle subcategories.

Table X.23: Databases Used to Estimate Off-Road Emissions

Databases Used to Estimate Off-Road Emissions	
Databases/Models	**Vehicle Subcategories**
ORION 2017, SORE 2020	Agriculture
ORION 2017, SORE 2020	Airport Ground Support
ORION 2017, SORE 2020	Cargo Handling Equipment
ORION 2017, SORE 2020	Construction and Mining
ORION 2017, SORE 2020	Industrial
SORE 2020	Lawn
ORION 2017, SORE 2020	Light Commercial
ORION 2017	Military Tactical Support
PC2014	Pleasure Crafts
ORION 2017	Portable Equipment
RV 2018	Recreational Vehicles
ORION 2017, SORE 2020	Transportation Refrigeration Unit

Source: CARB: ORION 2017, SORE 2020, PC2014 Pleasure Craft model, RV 2018 Recreational Vehicle model; EPIC, USD 2020.

The key inputs and 2016 GHG emissions are shown in Table X.24.

Table X.24: Key Inputs and 2016 Greenhouse Gas Emissions from Off-Road Transportation

Key Inputs and 2016 Greenhouse Gas Emissions from Off-Road Transportation	
Subcategories	**GHG Emissions (MMT CO_2e)**
Agriculture	0.010
Airport Ground Support	0.017
Cargo Handling Equipment	0.002
Construction and Mining	0.204
Industrial	0.097
Lawn	0.052
Light Commercial	0.071
Military Tactical Support	0.022
Pleasure Crafts	0.066
Portable Equipment	0.068
Recreational Vehicles	0.003
Transportation Refrigeration Unit	0.008
Total	**0.62**

Source: CARB: ORION 2017, SORE 2020, PC2014 Pleasure Craft model, RV 2018 Recreational Vehicle model; EPIC, USD 2020.

Difference from Previous 2012 Inventory

The previous 2012 inventory also relied on CARB's models to calculate emissions from off-road equipment. However, at that time, CARB had not yet developed ORION 2017, SORE 2020, or RV 2018 models; therefore, emissions were generated from either the 2007 or the 2011 OFFROAD model.

Method Used to Develop Emissions Projections

EPIC used the same models described in the previous section to generate emission projections for the subcategories, as shown in Table X.25.

Table X.25: Projected Greenhouse Gas Emissions from Off-Road Transportation

Projected Greenhouse Gas Emissions from Off-Road Transportation					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Agriculture (MMT CO_2e)	0.005	0.005	0.005	0.005	0.005
Airport Ground Support (MMT CO_2e)	0.02	0.02	0.02	0.03	0.03
Cargo Handling Equipment (MMT CO_2e)	0.004	0.005	0.006	0.006	0.006
Construction and Mining (MMT CO_2e)	0.25	0.28	0.30	0.33	0.35
Industrial (MMT CO_2e)	0.11	0.11	0.11	0.12	0.12
Lawn (MMT CO_2e)	0.060	0.061	0.063	0.065	0.066
Light Commercial (MMT CO_2e)	0.090	0.095	0.099	0.11	0.11
Military Tactical Support (MMT CO_2e)	0.022	0.022	0.022	0.022	0.022
Pleasure Crafts (MMT CO_2e)	0.074	0.079	0.085	0.097	0.104
Portable Equipment (MMT CO_2e)	0.081	0.090	0.099	0.121	0.133
Recreational Vehicles (MMT CO_2e)	0.004	0.004	0.004	0.005	0.005
Transportation Refrigeration Unit (MMT CO_2e)	0.010	0.010	0.011	0.012	0.012
Total (MMT CO_2e)	**0.72**	**0.79**	**0.83**	**0.91**	**0.95**

Source: CARB: ORION 2017, SORE 2020, PC2014 Pleasure Craft model, RV 2018 Recreational Vehicle model; EPIC, USD 2020.

Solid Waste

Emissions from solid waste are a result of biodegradable, carbon-bearing waste decomposing in largely anaerobic environments and producing landfill gas. The degradation process can take 5 to 50 years. Emissions from solid waste contribute to 2% of total emissions in the 2016 inventory and 4% in the 2050 projection. For this inventory, EPIC calculated the future emissions due to the waste disposed in 2016. Emissions due to waste-in-place are not calculated to be consistent with emissions included in the 2012 GHG inventory.

Method Used to Estimate 2016 Emissions

EPIC estimated the emissions from solid waste using method SW.4 from the ICLEI U.S. Community Protocol.[39] The emissions are based on the disposed waste in a given year, the characterization of the waste stream, and emissions factor of each type of waste. Because a waste characterization study for the entire region was not available, EPIC used the waste characterization studies from the cities of Chula Vista, Oceanside, and San Diego to estimate the waste composition in the region.[40] The solid waste emission

39 ICLEI: U.S. Community Protocol Appendix E, accessed in May 2020.

40 The City of Chula Vista and the City of Oceanside's waste characterization studies were provided by the jurisdictions. Personal communication. City of San Diego Waste Characterization Study.

factors, MT CO_2e per short ton of waste by type, are from the EPA Waste Reduction Model (WARM).[41] Table X.26 shows the waste composition derived and the corresponding emission factors.

Table X.26: Estimated San Diego Region Solid Waste Composition

Estimated San Diego Region Solid Waste Composition		
Type of Waste	Percentage of Total Composition*	Landfill Methane Without Recovery (MT CO_2e/Short Ton)
Paper	17%	2.12
Plastic	9.9%	0
Glass	1.9%	0
Metal	3.5%	0
Organics	40.4%	1.03
Electronics	0.8%	0
Inerts and Other	21.2%	0.07
Household Hazardous Waste	0.2%	0
Special Waste	2.9%	0
Mixed Residue	2.1%	0

* The composition was derived from the waste composition of the City of Chula Vista, City of Oceanside, and City of San Diego.

Source: EPIC, USD 2020.

The 2016 emissions from solid waste are provided in Table X.27.

Table X.27: Key Inputs and 2016 Greenhouse Gas Emissions from Solid Waste

Key Inputs and 2016 Greenhouse Gas Emissions from Solid Waste	
Total Waste Disposal (short tons)	3,317,216
Mixed Waste Emission Factor (MT CO_2e/short ton)*	0.79
Landfill Gas Capture Rate	0.75
Oxidation Rate	0.10
Total GHG Emissions (MMT CO_2e)	**0.59**

* Weighted average from Table X.26.

Source: EPIC, USD 2020.

41 U.S. EPA Waste Reduction Model (WARM) Version 15.

Difference from Previous 2012 Inventory

Because a waste characterization study was not available for the San Diego region, the previous 2012 inventory used the 2008 statewide waste characterization study. For the 2016 GHG inventory, EPIC used the more recent waste characterization studies available for the cities of Chula Vista, Oceanside, and San Diego.

Method Used to Develop Emissions Projections

EPIC projected the emissions, as shown in Table X.28, based on per capita waste disposal in 2016 and population growth.

Table X.28: Projected Greenhouse Gas Emissions from Solid Waste

Projected Greenhouse Gas Emissions from Solid Waste					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Total Waste Disposal (short tons)	3,470,848	3,552,485	3,620,348	3,719,685	3,746,073
Mixed Waste Emission Factor (MT CO_2e/short ton)	0.79	0.79	0.79	0.79	0.79
Landfill Gas Capture Rate	0.75	0.75	0.75	0.75	0.75
Oxidation Rate	0.10	0.10	0.10	0.10	0.10
Total GHG Emissions (MMT CO_2e)	**0.62**	**0.64**	**0.65**	**0.67**	**0.67**

Source: EPIC, USD 2020.

Water

The GHG emissions from energy associated with upstream supply and conveyance, and treatment of water account for 1% of total emissions in the 2016 inventory and none in the 2050 projection. This category does not include emissions associated with electricity used for water distribution and water end use (e.g., water heating at homes). The emissions from energy used for water distribution and water end use are captured in the electricity and natural gas categories, discussed in previous sections.

Method Used to Estimate 2016 Emissions

The San Diego County Water Authority (SDCWA) is the water wholesaler for the San Diego region. SDCWA imports raw and treated water on behalf of its 24 member agencies. The raw water sources, from the State Water Project and Colorado River, vary year by year depending on water availability; therefore, the energy needed to supply and convey water differs as well. The latest available upstream energy intensity, in kWh per acre-foot of water, is from the average of fiscal years 2013 and 2014 in the SDCWA 2015 Urban Water Management Plan. EPIC calculated the GHG emissions from upstream water supply by multiplying the water supplies with their respective energy intensities and the California average electricity GHG emission factor in 2016.[42] The upstream emissions are shown in Table X.29.[43]

Table X.29: 2016 Upstream Emissions from Water Supply

2016 Upstream Emissions from Water Supply		
Water Source	**Imported Treated Water**	**Imported Raw Water**
Water Demand (acre-feet)	138,312	282,726
Energy Intensity (kWh/acre-foot)*	1,862	1,817
California Average Electricity Emission Factor (lbs CO_2e/MWh)**	530	530
Upstream GHG Emissions (MT CO_2e)		185,411

* Includes water conveyance from the State Water Project and Colorado River to Metropolitan Water District and SDCWA system. The difference between energy intensity for treated and raw water is the water treatment energy intensity.

** eGRID 2016 CAMX subregion emission factor.

Source: EPIC, USD 2020.

[42] SDCWA 2016: Urban Water Management Plan 2015, Metropolitan Water District of Southern California, Urban Water Management Plan 2015. The Western Electricity Coordinating Council CAMX (eGRID Subregion) emission rate from eGRID was used as representative of the average California electricity emission rate for upstream electricity. U.S. EPA. eGRID 2016 Edition, released February 15, 2018, accessed June 29, 2018.

[43] 2016 water source and demand for each SDCWA member agency were provided by SDCWA staff to EPIC, October 23, 2018.

SDCWA has its own water treatment plant (WTP), Twin Oaks WTP, and many SDCWA member agencies have their own WTPs as well. Member agencies that do not have WTPs may purchase treated water from other member agencies or from SDCWA. For example, the City of San Diego and the City of Del Mar are member agencies of the SDCWA, but the City of San Diego provides water treatment service for the City of Del Mar. Local water treatment energy intensity depends on water sources, treatment level, capacity, and efficiency of the WTP. For example, brackish groundwater requires advanced treatment, such as reverse osmosis, to remove the salinity in the water, so its treatment has a higher energy intensity than surface water treatment with conventional methods. Table X.30 below shows the WTPs in the San Diego region, the water treated, and the associated electricity use for water treatment in 2016.[44] EPIC calculated the GHG emissions from water treatment by multiplying the electricity used for water treatment with SDG&E 2016 electricity GHG emission factor.

[44] Data were collected by EPIC from 2018 to 2020 for the development of SANDAG's 2016 and 2018 "ReCAP Snapshots" (greenhouse gas inventory and Climate Action Plan monitoring reports prepared for local jurisdictions).

Table X.30: 2016 Emissions from Local Water Treatment

2016 Emissions from Local Water Treatment				
Water Treatment Plant	**Plant Operator**	**Water Treated (Acre-Feet)**	**Water Treatment Energy Intensity (kWh/Acre-Foot)**	**Water Treatment Electricity Use (kWh)**
R.M. Levy WTP	Helix WD	42,767	58	2,493,844
R.E. Badger Filtration Plant	Santa Fe ID	12,685	44	558,346
Combined Miramar, Otay, and Alvarado WTP*	City of San Diego	163,823	56	9,151,144
Escondido-Vista WTP	Escondido + Vista ID	30,678	47	1,441,875
David C. McCollum WTP	Olivenhain MWD	21,301	142	3,018,745
Richard A. Reynolds Ground Water Desalination Facility	Sweetwater Authority	1,855	1,174	2,178,583
Robert A. Perdue WTP	Sweetwater	13,347	141	1,879,760
Lester J. Berglund WTP	City of Poway	10,329	208	2,150,666
Robert A. Weese WFP	City of Oceanside	11,878	29	348,546
Mission Basin Groundwater	City of Oceanside	2,997	1,257	3,766,499
Twin Oaks Valley WTP	SDCWA	79,538	33	2,661,602
Carlsbad Desalination Plant**	SDCWA	45,107	4,397	198,335,919
Total Water Treatment Electricity Use (kWh)				227,985,529
SDG&E Electricity Emission Factor (lbs CO_2e/MWh)				527
Transmission and Distribution Loss Factor				1.082
Local Treatment GHG Emissions (MT CO_2e)				58,925

ID: irrigation district; WD: water district; WFP: water filtration plant; WTP: water treatment plant.

* The electricity use and energy intensity include both water treatment and conveyance from nearby reservoirs for City of San Diego WTPs and both water extraction and treatment for Sweetwater Authority's brackish water desalination plant. The data associated with water treatment cannot be separated out.

** The water treated at the plant includes SDCWA wholesale water and local supply for individual SDCWA member agencies that have separate contracts with the plant. The energy intensity is the high efficiency estimate from the Plant's Environmental Impact Report (2008).

Source: EPIC, USD 2020.

Combining the upstream and local emissions, the total 2016 emissions from water are shown in Table X.31.

Table X.31: 2016 Greenhouse Gas Emissions from Water Supply, Treatment, and Distribution

2016 Greenhouse Gas Emissions from Water Supply, Treatment, and Distribution	
Upstream GHG Emissions (MT CO_2e)	185,411
Local Treatment GHG Emissions (MT CO_2e)	58,925
Total (Upstream + Local) GHG Emissions (MT CO_2e)	244,337
Total (Upstream + Local) GHG Emissions (MMT CO_2e)	**0.24**

Source: EPIC, USD 2020.

Difference from Previous 2012 Inventory

The methods to calculate water emissions are different. Due to data availability, the 2012 GHG emissions from water use were based on default per capita water production and Southern California–specific water-energy intensities. 2016 emissions from water use were based on region-specific water production data and specific treatment facility information.

Method Used to Develop Emissions Projections

To project emissions for the water category, EPIC estimated the impact of federal and state polices and regulations on reducing the electricity emission factor (increasing renewable or zero-carbon electricity) and increasing water efficiency, respectively.

As discussed in the *Electricity* section, all retail electricity suppliers must meet the RPS requirement of 60% renewable electricity by 2030 and 100% renewable or zero-carbon electricity by 2045. EPIC assumed all retail electricity providers that provide electricity for water supply and treatment will meet the 2030 and 2045 RPS targets. The renewable or GHG-free content and emission factors for water-related electricity use are shown in Table X.32.

Table X.32: Projected Renewable or Greenhouse Gas–Free Content and Emission Factors for Water-Related Electricity Use

Projected Renewable or Greenhouse Gas–Free Content and Emission Factors for Water-Related Electricity Use					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Projected Renewable or GHG-Free Content (%)*					
California Average	47%	60%	73%	100%	100%
San Diego Region	47%	60%	73%	100%	100%
Projected Electricity Emission Factor (lbs CO_2e/MWh)**					
California Average	493	370	249	—	—
San Diego Region	493	370	249	—	—

Retail electricity suppliers in the San Diego region may be SDCP, CEA, SDG&E, or others. SDG&E's projected renewable content and emission factors are used as a conservative approach.

* Estimated based on 2016 California average and SDG&E renewable content, and SB 100 RPS targets.

** Calculated based on 2016 SDG&E bundled electricity emission factor of 527 lbs CO_2e/MWh and 43% renewable content.

Source: EPIC, USD 2020.

SDCWA's preliminary 2020 Urban Water Management Plan estimates the long-range water demand in its service area through 2045. The water demand forecasts include a baseline demand forecast (based on the SANDAG projected growth forecast, local weather data, historical water use, and retail rates) and a long-range demand forecast with additional water conservation savings. The additional water conservation savings include both "active" program savings (from implementation of water conservation programs) and "passive" code-based water savings (future savings from appliance standards, plumbing code changes, and updated Model Water Efficient Landscape Ordinances).[45] EPIC applied the long-range demand forecast rate of increase to the 2016 water demand to be consistent with the projection methods in other emissions categories. As no forecast is available after 2045, EPIC used the 2040–2045 annual demand increases as the 2045–2050 annual increases. Assuming the water-energy intensities are fixed, the projected emissions are shown in Table X.33.

[45] SDCWA Water Planning and Environmental Committee November 4, 2020, Meeting: Report on Preparation of Draft 2020 Urban Water Management Plan. (Presentation), accessed January 3, 2021.

Table X.33: Projected Greenhouse Gas Emissions from Water

Projected Greenhouse Gas Emissions from Water					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Projected Upstream Emissions					
Imported Treated Water (acre-feet)	170,707	177,593	183,634	193,411	193,411
Imported Raw Water (acre-feet)	348,945	363,020	375,368	395,354	406,000
California Average Emission Factor (lbs CO_2e/MWh)	493	370	249	—	—
Upstream Emissions (MT CO_2e)*	212,754	166,002	115,863	—	—
Projected Local Emissions					
Water Treated at Local Water Treatment Plants (acre-feet)	538,496	560,218	579,273	610,115	626,544
San Diego Region Emission Factor (lbs CO_2e/MWh)	493	370	249	—	—
Local Emissions (MT CO_2e)**	68,048	53,095	37,058	—	—
Projected Total Emissions					
Total (Upstream + Local) Emissions (MT CO_2e)	280,803	219,097	152,921	—	—
Total Emissions (MMT CO_2e)	**0.28**	**0.22**	**0.15**	**—**	**—**

* Assume upstream energy intensities 1,862 kWh/acre-foot for imported treated water and 1,817 kWh/acre-foot for imported untreated water remain unchanged (Table X.29).

** Assume energy intensities at local water treatment plants remain unchanged (Table X.30).

Source: EPIC, USD 2020.

Civil Aviation

The GHG emissions from commercial aviation operations account for 1% of total emissions in the 2016 inventory and 2% in the 2050 projection. The San Diego International Airport (SAN) and McClellan-Palomar Airport (CRQ) are the only airports in the San Diego region in 2016 with scheduled commercial flights services, while other airports operate on a private and on-demand basis.[46] Because 99% of commercial passengers in the San Diego region are covered by SAN and CRQ, this category does not include the GHG emissions associated with aviation operations at other municipal airports in the San Diego region.[47] GHG emissions in this category are from combustion of jet fuel and aviation gasoline used by commercial aircrafts.

[46] Airports with scheduled commercial flights follow Federal Aviation Administration (FAA)'s FAR Part 139 rules. On-demand basis refers to aviation operators allowed under FAA rules to accept paying passengers (FAR Part 135 operators).

[47] FAA: Passenger Boarding (Enplanement) and All-Cargo Data for U.S. Airports, CY 2016. Airports included are SAN, CRQ, Miramar MCAS, North Island NAS, Montgomery-Gibbs, Brown Field, and Gillespie Field.

Method Used to Estimate 2016 Emissions

EPIC used the aircraft emissions reported in the SAN 2016 GHG Emissions Inventory (SAN GHG Inventory)—developed by the San Diego County Regional Airport Authority—and CRQ 2016 Emissions Inventory—developed for the CRQ Master Plan Program Environmental Impact Report (PEIR). The aircraft emissions in the SAN GHG Inventory are calculated based on the Airport GHG Emissions Management Guidance Manual and include emissions from aircraft start up, take off, and up to mixing height (3,000 feet).[48] The aircraft emissions in CRQ 2016 Emissions Inventory include emissions from fuel combustion and emissions from auxiliary power units.[49]

The 2016 aircraft emissions were 213,353 (0.2 MMT CO_2e), with 95% from SAN aircraft emissions and 5% from CRQ aircraft emissions.

Difference from Previous 2012 Inventory

In both inventories, emissions from the SAN GHG Inventories were used directly. However, the 2016 Airport GHG Inventory is calculated with Airports Council International's Airport Carbon and Emissions Reporting Tool, which no longer includes the emissions from aircrafts during cruise (above mixing height: 3,000 feet). The 2012 Airport GHG Inventory included emissions from the entire flight. In addition, aircraft emissions from CRQ were added to 2016 GHG emissions.

Method Used to Develop Emissions Projections

To project emissions for the civil aviation category, EPIC applied the rate of increase of the projected passengers served at the SAN to the 2016 aircraft emissions. In 2016, SAN served a total of 20,729,353 passengers.[50] The draft SAN Development Plan projects the number of passengers served with the proposed Terminal 1 replacement and Terminal 2 modification. Under the constrained demand scenario, the SAN Development Plan projects an average increase of 1.6% per year in passengers from 2018 to 2050.[51] EPIC applied the 1.6% annual increase to the SAN aircraft emissions in 2016. For CRQ, the projected 2036 aircraft emissions under proposed CRQ Master Plan are used directly and kept fixed through 2050.[52] The projected emissions are shown in Table X.34.

---

48 San Diego County Regional Airport Authority: 2016 Greenhouse Gas Emissions Inventory (October 16, 2018), provided by Airport Authority staff to EPIC, August 7, 2018.

49 CRQ Master Plan Update PEIR: Appendix H – Climate Change Technical Report (2018).

50 San Diego International Airport: Air Traffic Report, January 2017, accessed December 23, 2020.

51 San Diego County Regional Airport Authority: Airport Development Plan Recirculated Draft EIR (September 2019), accessed January 10, 2021.

52 CRQ Master Plan Update PEIR: Appendix H – Climate Change Technical Report (2018).

Table X.34: 2016 Greenhouse Gas Emissions and Projected Greenhouse Gas Emissions from Civil Aviation

2016 Greenhouse Gas Emissions and Projected Greenhouse Gas Emissions from Civil Aviation						
Year	**2016**	**2025**	**2030**	**2035**	**2045**	**2050**
SAN Airport Total Passengers*	20,729,353	27,736,698	30,027,785	32,508,118	38,100,345	41,247,483
SAN Passengers Increase Compared with 2016	0%	34%	45%	57%	84%	99%
SAN GHG Emissions (MT CO_2e)	202,422	270,849	293,221	317,442	372,050	402,781
CRQ GHG Emissions (MT CO_2e)	10,931	18,204	22,244	26,284	27,093	27,093
Total GHG Emissions (MT CO_2e)	213,353	289,052	315,465	343,726	399,142	429,874
Total GHG Emissions (MMT CO_2e)	**0.21**	**0.29**	**0.32**	**0.34**	**0.40**	**0.43**

SAN: San Diego International Airport; CRQ: McClellan-Palomar Airport.

* 2016 total passengers are from the San Diego International Airport 2017 Air Traffic Report, and the rest are based on an annual increase of 1.6%.

Source: EPIC, USD 2021.

Rail

The rail category includes GHG emissions from both passenger and freight rail resulting from the combustion of fuels in internal combustion engines. Emissions from rail contribute to 0.4% of total emissions in the 2016 inventory and 1% in the 2050 projection.

Method Used to Estimate 2016 Emissions

Detailed activity or fuel consumption data for rail were not available for the San Diego region. EPIC scaled the emissions from the CARB statewide inventory to the San Diego region, based on the ratio of 2016 County Business Pattern establishments for support activities for rail transportation to that of the state.[53]

Because the rail category in CARB's statewide inventory is not separated into freight and passenger rail subcategories, EPIC used the number of support establishments for rail in the San Diego region to capture both freight and passenger rail activities. However, it may not represent the exact ratio of all rail in the region compared to the state. The most recent 2018 County Business Pattern data do not show any data on support establishments for rail transportation for the San Diego region; therefore, the method used in this appendix may be limited. Table X.35 shows the key inputs and 2016 GHG emissions from rail.

Table X.35: Key Inputs and 2016 Greenhouse Gas Emissions from Rail

Key Inputs and 2016 Greenhouse Gas Emissions from Rail	
Support Activities for Rail Transportation in California	78
Support Activities for Rail Transportation in San Diego Region	4
Total Rail Emissions in California (MMT CO_2e)	2.17
Total Rail Emissions in San Diego (MMT CO_2e)	**0.11**

Support Activities for Rail Transportation: NAICS 4882. Industries under NAICS 4882 provide services that support rail transportation.

Source: EPIC, USD 2020.

Difference from Previous 2012 Inventory

Methods to estimate emissions from rail are the same in both the 2012 and 2016 inventories.

Method Used to Develop Emissions Projections

EPIC projected the emissions from rail based on the SANDAG jobs forecast, as shown in Table X.36.

[53] CARB: CARB Greenhouse Gas Emission Inventory – Query Tool, accessed on October 25, 2020. U.S. Census Bureau: 2016 County Business Patterns, accessed on October 25, 2020. The NAICS Code for rail transportation support activities is 4882.

Table X.36: Projected Greenhouse Gas Emissions from Rail

Projected Greenhouse Gas Emissions from Rail					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Total GHG Emissions (MMT CO_2e)	0.17	0.19	0.19	0.20	0.20

Source: EPIC, USD 2020.

Wastewater

The GHG emissions from domestic wastewater treatment account for 0.3% of total emissions in the 2016 inventory and 0.5% in the 2050 projection. This category presents emissions from community-generated wastewater treated at centralized wastewater treatment plants and on-site septic systems. Emissions associated with the energy used to collect and treat wastewater are not included in this category but are included in the electricity and natural gas category.

Method Used to Estimate 2016 Emissions

In 2019, SANDAG, in collaboration with local jurisdictions, prepared the 2016 Regional Climate Action Planning Framework (ReCAP) Snapshots to assist local jurisdictions with monitoring community-wide GHG emissions and Climate Action Plan (CAP) implementation.[54] EPIC calculated the 2016 community-wide GHG emissions inventories for 16 (out of 19) jurisdictions in the San Diego region and used the wastewater emissions from these 16 GHG inventories directly in this category.

The City of Coronado postponed preparation of a ReCAP Snapshot due to the ongoing CAP development; however, 2016 wastewater flow was collected during the data-collection process. The GHG emissions shown in Table X.37 for Coronado include wastewater flow from military bases in Coronado to the Point Loma Wastewater Treatment Plant (WWTP). Depending on the boundary determined in the future Coronado CAP, the wastewater emissions estimated here may differ from those calculated under the CAP.

The City of San Diego and the unincorporated County of San Diego (the County) report community-wide GHG emissions separately under their own CAP monitoring processes. The 2016 wastewater emissions from the City of San Diego are taken directly from its 2019 CAP Annual Report.[55] For the County, EPIC estimated the 2016 wastewater emissions using its 2014 (CAP baseline year) wastewater emissions and population increase.[56]

54 SANDAG: Climate Action. November 2019 ReCAP Snapshots (with 2016 GHG Emissions Inventories).
55 City of San Diego CAP: 2019 Annual Report Appendix (2020), accessed November 2, 2020.
56 County of San Diego CAP Appendix A: 2014 Greenhouse Gas Emissions Inventory and Projections (2017), accessed May 20, 2020.

The key inputs and 2016 wastewater emissions are show in Table X.37.

Table X.37: Key Inputs and 2016 Greenhouse Gas Emissions from Wastewater

Key Inputs and 2016 Greenhouse Gas Emissions from Wastewater	
Local Jurisdiction	**2016 Wastewater Emissions (MT CO_2e)**
Carlsbad	2,972
Chula Vista	2,577
Coronado	260
Del Mar	87
El Cajon	1,161
Encinitas	1,916
Escondido	4,986
Imperial Beach	353
La Mesa	734
Lemon Grove	260
National City	656
Oceanside	5,751
Poway	1,140
San Diego*	21,257
San Marcos	2,915
Santee	584
Solana Beach	619
Vista	3,207
Unincorporated County of San Diego**	21,583
Total	73,014
Total (MMT CO_2e)	**0.07**

* 2016 emissions reported in the City of San Diego CAP 2019 Annual Report.

** Estimated based on 2014 wastewater emissions reported in the County of San Diego CAP Appendix A (21,183 MT CO_2e), 2014 population (498,159), and 2016 population (507,555).

All wastewater emissions are from SANDAG November 2019 ReCAP Snapshots (with 2016 GHG Emissions), except City of San Diego and County of San Diego.

Source: SANDAG 2019, EPIC, USD 2020.

Difference from Previous 2012 Inventory

The methods to calculate wastewater emissions are different from those used in the previous 2012 inventory. Due to data availability, the 2012 wastewater emissions were based on a default per capita wastewater production in California and Point Loma WWTP's wastewater emission factor. The 2016 wastewater emissions were based on jurisdictional wastewater flow data and specific wastewater treatment facility information.

Method Used to Develop Emissions Projections

To project emissions for the wastewater category, EPIC applied the population rate of increase from 2016 to 2050 to the 2016 wastewater emissions. The projected emissions are shown in Table X.38.

Table X.38: Projected Greenhouse Gas Emissions from Wastewater

Projected Greenhouse Gas Emissions from Wastewater						
Projection Year	**2016**	**2025**	**2030**	**2035**	**2045**	**2050**
San Diego Region Population*	3,287,280	3,470,848	3,552,485	3,620,348	3,719,685	3,746,073
Population Increase Compared with 2016 (%)	—	6%	8%	10%	13%	14%
Total GHG Emissions (MT CO_2e)	73,014	77,091	78,904	80,412	82,618	83,204
Total GHG Emissions (MMT CO_2e)	**0.07**	**0.08**	**0.08**	**0.08**	**0.08**	**0.08**

* 2016 population data are estimates, the rest are from SANDAG Series 14 Growth Forecast, as shown in Table X.2.

Source: EPIC, USD 2021.

Agriculture

GHG emissions from livestock (from enteric fermentation and manure management) are included in this category. Enteric fermentation is a microbial fermentation process that occurs in the stomach of ruminant animals, producing CH_4 that is released through flatulence and eructation. Manure management is the process by which manure is stabilized or stored. CH_4 and N_2O emissions result from livestock manure, and the amount of gas produced depends on the manure management system involved. The agriculture category contributes to 0.2% of total emissions in the 2016 inventory and 0.3% in the 2050 projection.

Method Used to Estimate 2016 Emissions

EPIC followed the ICLEI *U.S. Community Protocol for Emissions from Domestic Animal Production within a Community* (A.1 and A.2) to calculate the emissions from agriculture.[57] Method A.1 addresses enteric fermentation from livestock production. CH_4 emissions due to enteric fermentation are derived from the livestock population and emission factors for each animal type. Method A.2 addresses emissions from manure management. Emissions from manure management are derived from data on animal populations, animal characteristics, and manure management practices. Method A.2 is broken up into three subcategories, including CH_4 emissions from manure management (A.2.1), direct N_2O emissions from manure management (A.2.3), and indirect N_2O emissions from manure management (A.2.4).

All the emission factors and other factors used for the calculations were taken from the ICLEI protocol. Table X.39 shows the factors used to calculate the agriculture emissions.

Table X.39: Factors Used to Calculate Greenhouse Gas Emissions from Agriculture

Factors Used to Calculate Greenhouse Gas Emissions from Agriculture							
	Dairy Cattle	**Other Cattle, Including Calves**	**Beef Cattle**	**Sheep**	**Goats**	**Swine**	**Horses**
Methane Emissions from Enteric Fermentation (A.1)							
Enteric Fermentation Emission Factor (kg CH_4/head/year)	147	54	100	8	5	1.5	18
Methane Emissions from Manure (without anaerobic digester) (A.2.1)							
Percentage Dry Lot	0	0.11	1	0.5	0.5	0	0.5
Percentage Pasture	0	0	0	0.5	0.5	0.2	0.5
Percentage Liquid Slurry	0.2	0.09	0.01	0	0	0.07	0
Percentage Daily Spread	0.1	0.01	0	0	0	0	0
Percentage Solid Storage	0.09	0	0	0	0	0	0
Percentage Anaerobic Lagoon	0.6	0.21	0	0	0	0.43	0
Percentage Dip Pit	0	0.58	0	0	0	0.27	0
Volatile Solid (VS) (kg/animal/year)	2,025	1,252	1,259	0	0	0	0

57 ICLEI: U.S. Community Protocol for Emissions from Domestic Animal Production within a Community, accessed August 3, 2020.

Factors Used to Calculate Greenhouse Gas Emissions from Agriculture							
	Dairy Cattle	Other Cattle, Including Calves	Beef Cattle	Sheep	Goats	Swine	Horses
Average VS (kg/day/1,000 kg animal mass)	0	0	0	8.3	9.5	5.4	6.1
Typical Animal Mass	0	0	0	25	64	39	450
Max CH_4 Producing Capacity per Pound of Manure (m^3 kg VS)	0.24	0.17	0.33	0.36	0.17	0.48	0.33
Methane Conversion Factor Pasture	0.015	0	0	0.015	0.015	0.015	0.015
Methane Conversion Factor Dry Lot	0	0.015	0.015	0.015	0.015	0	0.015
Methane Conversion Factor Liquid Slurry	0.34	0.35	0.43	0	0	0.33	0
Methane Conversion Factor Daily Spread	0.005	0.005	0	0	0	0	0
Methane Conversion Factor Solid Storage	0.04	0	0	0	0	0.04	0
Methane Conversion Factor Anaerobic Lagoon	0.73	0.75	0	0	0	0.73	0
Methane Conversion Factor Dip Pit	0	0.35	0	0	0	0.33	0
Direct Nitrous Oxide Emissions from Manure (A.2.3)							
Daily Rate of Kjeldahl Nitrogen Excreted (kg N/animal/year)	156	54.7	52.3	0.45	0.45	0.54	0.25
Direct N_2O Emission Factor Dry Lot	0	0.03	0.02	0.03	0.03	0	0.03
Direct N_2O Emission Factor Pasture	0	0	0	0	0	0	0
Direct N_2O Emission Factor Daily Spread	0	0	0	0	0	0	0
Direct N_2O Emission Factor Solid Storage	0.005	0	0	0	0	0.005	0
Direct N_2O Emission Factor Liquid/Slurry	0.005	0.08	0.005	0	0	0.08	0

Factors Used to Calculate Greenhouse Gas Emissions from Agriculture							
	Dairy Cattle	Other Cattle, Including Calves	Beef Cattle	Sheep	Goats	Swine	Horses
Direct N_2O Emission Factor Dip Pit	0	0.002	0	0	0	0.002	0
Direct N_2O Emission Factor Anaerobic Lagoon	0	0	0	0	0	0	0
Indirect Nitrous Oxide Emissions from Manure (A.2.4)							
Frac Gas, Pasture*	0	0	0	0	0	0	0
Frac Gas, Liquid/Slurry	26	26	26	0	0	26	0
Frac Gas, Daily Spread	10	10	0	0	0	0	0
Frac Gas, Dry Lot	0	23	23	23	23	0	23
Frac Gas, Solid Storage	27	0	0	0	0	45	0
Frac Gas, Anerobic Lagoon	43	43	0	0	0	58	0
Frac Gas, Dip Pit	0	24	0	0	0	34	0
Frac Runoff/Leach, Pasture**	0	0	0	0	0	0	0
Frac Runoff/Leach, Daily Spread	0	0	0	0	0	0	0
Frac Runoff/Leach, Solid Spread	0	0	0	0	0	0	0
Frac Runoff/Leach, Liquid/Slurry	0.8	0.8	0	0	0	0.8	0
Frac Runoff/Leach, Anaerobic Lagoon	0.8	0.8	0	0	0	0.8	0
Frac Runoff/Leach, Dry Lot	0	3.9	3.9	3.9	3.9	0	0
Frac Runoff/Leach, Dip Pit	0	0	0	0	0	0	0

* Frac Gas – Nitrogen lost through volatilization.

** Frac Runoff/Leach – Nitrogen lost through runoff and leaching.

Source: ICLEI 2013; EPIC, USD 2020.

Table X.40 shows the GHG emissions from agriculture.

Table X.40: 2016 GHG Emissions from Agriculture

2016 GHG Emissions from Agriculture	
Animal Population (Head)	
Dairy Cattle	1,800
Other Cattle, Including Calves	5,400
Beef Cattle	3,700
Sheep	928
Goats	2,700
Swine	1,220
Horses	6,813
CH_4 Emission from Enteric Fermentation (A.1)	
Dairy Cattle Enteric Fermentation Emissions (MT CO_2e)	6,615
Other Cattle Enteric Fermentation Emissions (MT CO_2e)	7,290
Beef Cattle Enteric Fermentation Emissions (MT CO_2e)	9,250
Sheep Enteric Fermentation Emissions (MT CO_2e)	186
Goats Enteric Fermentation Emissions (MT CO_2e)	338
Swine Enteric Fermentation Emissions (MT CO_2e)	46
Horses Enteric Fermentation Emissions (MT CO_2e)	3,066
Total CO_2e Emissions from Enteric Fermentation (MMT CO_2e)	0.027
CH_4 Emissions from Manure (Without Anaerobic Digester) (A.2.1)	
CH_4 Emissions from Volatile Solids (VS) Excreted from Beef Cattle (MT CO_2e)	491
CH_4 Emissions from VS Excreted from Dairy Cattle (MT CO_2e)	7,385
CH_4 Emissions from VS Excreted from Other Cattle (MT CO_2e)	7,486
CH_4 Emissions from VS Excreted from Swine (MT CO_2e)	321
CH_4 Emissions from VS Excreted from Sheep (MT CO_2e)	6.3
CH_4 Emissions from VS Excreted from Goats (MT CO_2e)	22
CH_4 Emissions from VS Excreted from Horses (MT CO_2e)	3076
Total CH_4 Emissions from VS Excreted from Domesticated Animals (MMT CO_2e)	0.019
Direct N_2O Emissions from Manure (A.2.3)	
Direct N_2O Emissions from Beef Cattle	1,817
Direct N_2O Emissions from Dairy Cattle (MT CO_2e)	191
Direct N_2O Emissions from Other Cattle (MT CO_2e)	1,613
Direct N_2O Emissions from Swine (MT CO_2e)	28
Direct N_2O Emissions from Sheep (MT CO_2e)	27

2016 GHG Emissions from Agriculture	
Direct N_2O Emissions from Goats (MT CO_2e)	200
Direct N_2O Emissions from Horses (MT CO_2e)	1,967
Total Direct N_2O Emissions from Manure (MMT CO_2e)	0.006
Indirect N_2O Emissions from Manure (A.2.4)	
Indirect N_2O Emissions from Beef Cattle	237
Indirect N_2O Emissions from Dairy Cattle (MT CO_2e)	389
Indirect N_2O Emissions from Other Cattle (MT CO_2e)	526
Indirect N_2O Emissions from Swine (MT CO_2e)	524
Indirect N_2O Emissions from Sheep (MT CO_2e)	2.9
Indirect N_2O Emissions from Goats (MT CO_2e)	17
Indirect N_2O Emissions from Horses (MT CO_2e)	151
Total Indirect N_2O Emissions from Manure (MMT CO_2e)	0.002
Total GHG Emissions from Agriculture (MMT CO_2e)	**0.05**

Source: EPIC, USD 2020.

Difference from Previous 2012 Inventory

Methods to estimate emissions from agriculture are the same in both the 2012 and 2016 inventories.

Method Used to Develop Emissions Projections

While the previous inventory used a logarithmic decay function to project the emissions out to 2050, the current inventory used a constant value for the years 2020–2050. Because livestock population in the San Diego region does not have a definitive growth pattern, a constant number was used for the emission projections. EPIC projected both enteric fermentation and manure management emission estimates to 2050 (Table X.41), based on the average 2017–2019 cattle population, which was kept constant for the years 2020–2050.[58]

Table X.41: Projected Emissions from Agriculture

Projected Emissions from Agriculture					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Total GHG Emission (MMT CO_2e)	0.06	0.06	0.06	0.06	0.06

Source: EPIC, USD 2020.

58 County of San Diego: 2017 County of San Diego Crop Statistics and Annual Report, 2018 County of San Diego Crop Statistics and Annual Report, 2019 County of San Diego Crop Statistics and Annual Report.

Marine Vessels

The GHG emissions from marine vessels in the San Diego region are largely attributed to the Port of San Diego, which serves as a transshipment facility for San Diego, Orange, Riverside, San Bernardino, and Imperial Counties, as well as northern Baja California and Arizona. The GHG emissions from marine vessels account for 0.2% of total emissions in the 2016 inventory and 0.5% in the 2050 projection.

The emissions are from the following two subcategories:

- **Ocean-Going Vessels (OGV):** These include auto carriers, bulk carriers, passenger cruise vessels, general cargo vessels, refrigerated vessels (reefers), roll-on roll-off vessels, and tankers for bulk liquids.
- **Commercial Harbor Craft (CHC):** These include tugboats, towboats, pilot boats, work boats, ferries, and sports and commercial fishing vessels.

The emissions from OGV or CHC beyond the Port of San Diego's landside and waterside boundary (24 nautical miles from the coastline) are not included in the 2016 inventory.

Method Used to Estimate 2016 Emissions

EPIC used the OGV and CHC emissions reported in the Port of San Diego 2016 Maritime Air Emissions Inventory.[59] The 2016 emissions are shown in Table X.42.

Table X.42: 2016 Greenhouse Gas Emissions from Marine Vessels

2016 Greenhouse Gas Emissions from Marine Vessels	
Vessel Type	**2016 Emissions**
OGV (MT CO_2e)	22,500
CHC (MT CO_2e)	25,500
Total GHG Emissions (MT CO_2e)	48,000
Total GHG Emissions (MMT CO_2e)	**0.05**

Source: Port of San Diego, 2018.

Difference from Previous 2012 Inventory

In both 2012 and 2016 inventories, emissions from the Port of San Diego Maritime Air Emissions Inventories are used directly. Port-related operations data were refined in the 2016 inventory; however, emission boundaries and methods to calculate emissions are the same.

[59] Port of San Diego 2016 Maritime Air Emissions Inventory (2018), accessed May 8, 2020. Other emissions from the 2016 Port of San Diego inventory, e.g., cargo handling equipment, locomotives, and on-road vehicles, are included in "Other categories" of this regional inventory.

Method Used to Develop Emissions Projections

To project emissions for the marine vessel category, EPIC used the projected OGV and CHC emissions in the San Diego region in the CARB ORION database.[60] The emissions from the ORION database include the impacts of adopted rules and regulations in each subcategory, as shown below:

- OGV Clean Fuel Regulation (beginning in 2009) and North American Emission Control Area (beginning in 2015)
- OGV At-Berth Regulation (2007) and proposed regulation (implementation through 2029)
- CHC Regulation (2007, amended in 2010, fully implemented by 2022)[61]

Because the boundaries are different for the OGV and CHC emissions reported by the Port of San Diego 2016 maritime air emissions and the ORION database, EPIC applied the rate of increase of the projected emissions in the ORION database to the 2016 Port of San Diego–calculated maritime emissions. The projected emissions are shown in Table X.43.

Table X.43: 2016 Greenhouse Gas Emissions and Projected Greenhouse Gas Emissions from Marine Vessels

2016 Greenhouse Gas Emissions and Projected Greenhouse Gas Emissions from Marine Vessels

Projection Year	2016	2025	2030	2035	2045	2050
OGV Emissions from ORION Compared with 2016*	—	31%	52%	75%	128%	156%
OGV Emissions	22,500	29,525	34,204	39,264	51,412	57,501
CHC Emissions from ORION Compared with 2016*	—	0.4%	1%	0.4%	–1%	–2%
CHC Emissions	25,500	25,606	25,646	25,613	25,211	24,867
Total GHG Emissions (MT CO_2e)	48,000	55,131	59,850	64,877	76,623	82,368
Total GHG Emissions (MMT CO_2e)	**0.05**	**0.06**	**0.06**	**0.06**	**0.08**	**0.08**

* San Diego region only. Emissions in ORION database are reported in tons per day.

Source: EPIC, USD 2020.

60 CARB: Emissions Inventory Offroad Emissions, accessed December 23, 2020.

61 CARB: Mobile Source Emissions Inventory – Off-Road Diesel Equipment Documentation.

Soil Management

Emissions from synthetic fertilizer use and crop residue or soil management contribute to 0.2% of total emissions in the 2016 inventory and 0.2% in the 2050 projection. The emissions are broken into two subcategories: farm emissions and non-farm emissions. The farm emissions account for the emissions due to agricultural soil management activities, such as synthetic fertilizers used for cultivation purposes to enhance the soil's nutrients and emissions due to crop residue. The non-farm emissions account for synthetic fertilizers used for residential or commercial purposes.

Farm emissions due to agricultural synthetic fertilizer use include direct N_2O emissions, indirect N_2O emissions, and CO_2 emissions from urea and lime application. The non-farm emissions only include direct N_2O and indirect N_2O emissions. The N_2O emissions from crop residues are due to the nitrogen content in the residue.

Method Used to Estimate 2016 Emissions

EPIC followed the IPCC method to calculate the direct and indirect N_2O and CO_2 emissions from managed soils.[62] The IPCC method calculates emissions from manure management, fertilizer application, and agricultural activities. Because the emissions from manure management are accounted for in the agriculture category, this section does not include these emissions.

To calculate the direct and indirect N_2O emissions from fertilizer applications for both farm and non-farm activities, EPIC multiplied the tonnage used by the nitrogen content of each synthetic fertilizer.[63] The nitrogen content of each fertilizer is based on the specific chemical content.[64] If the specific chemical content of a fertilizer is not given, code 97 fertilizer with a 25-15-17 Nitrogen-Phosphorous-Potassium (NPK) composition is used.

The farm use soil management has N_2O emissions from crop residue and from crop burning activities. Because the San Diego region does not have agricultural burning activities in 2016, EPIC only considered the emissions due to crop residue. Among the crops that have nitrogen content in their residue, only oats/hay are grown in the San Diego region. EPIC calculated the emissions from crop residue using the total nitrogen content in the crop residue based on the acres of crop cultivated.[65] The CO_2 emissions from urea application and from liming are based on the total quantities of urea and lime applied and their respective emission factors.[63] Table X.44 shows the key inputs and results for the soil management emissions.

---

62 IPCC: N_2O Emissions from Managed Soils and CO_2 Emissions from Urea and Lime Application, accessed on August 2, 2020.

63 California Department of Food & Agriculture: 2016 Fertilizing Materials Tonnage Report, accessed on August 3, 2020.

64 International Fertilizer Association: Fertilizer Converter, accessed on August 3, 2020. This database provides information on the nitrogen content percentage by weight of a given fertilizer.

65 California Department of Agriculture Weights & Measures: 2016 County of San Diego Crop Statistics and Annual Report, accessed on August 4, 2020.

Table X.44: Key Inputs and 2016 Greenhouse Gas Emissions from Soil Management

Key Inputs and 2016 Greenhouse Gas Emissions from Soil Management	
Total Nitrogen in Farm Use Synthetic Fertilizers (tons)	3,013
Total Nitrogen in Non-Farm Use Synthetic Fertilizers (tons)	5,247
N_2O Emitted per Unit of Nitrogen (kg N_2O-N/kg N)	0.01
N_2O Emitted per Unit of Nitrogen Volatilized (kg N-N_2O/kg NH_3- N + NO_x-N volatilized)	0.01
N_2O Emitted per Unit of Nitrogen Leached/Runoff (kg N_2O-N/kg N leaching/runoff)	0.0075
Total Area of Oats Harvested (acres)	2,100
Total Nitrogen in Crop (Oats/Hay) Residue (kg N)	7,990
Amount on Lime Applied (tons)	216
Carbon Content of Lime (ton C/ton of lime)	0.125
Amount of Urea Applied (tons)	559
Carbon Content of Urea (ton C/ton of urea)	0.2
Direct N_2O Emissions from Farm Activities – Synthetic Fertilizers and Crop Residue (MMT of CO_2e)	0.013
Direct N_2O Emissions from Non-Farm Activities – Synthetic Fertilizer (MMT of CO_2e)	0.022
Indirect N_2O Emissions from Farm Activities – Synthetic Fertilizers and Crop Residue (MMT of CO_2e)	0.004
Indirect N_2O Emissions from Non-Farm Activities – Synthetic Fertilizer (MMT of CO_2e)	0.007
CO_2 Emissions from Farm Urea Applications (MMT CO_2e)	4×10^{-4}
CO_2 Emissions from Farm Lime Applications (MMT CO_2e)	1×10^{-4}
Total Farm Emissions (MMT CO_2e)	0.02
Total Non-Farm Emissions (MMT CO_2e)	0.03
Total GHG Emissions from Soil Management Sector (MMT CO_2e)	**0.05**

Source: County of San Diego 2016; International Fertilizer Association IPCC 2006; EPIC, USD 2020.

Difference from Previous 2012 Inventory

The previous 2012 inventory did not include emissions from soil management activities (fertilizer application and crop residue).

Method Used to Develop Emissions Projections

Direct and indirect N_2O and CO_2 emissions were projected to 2050 using the Microsoft Excel GROWTH function. EPIC calculated the emissions for the years 2016–2019 using the available data for oats harvested, fertilizer use, and the IPCC emission factors and projected the activity data out to 2050 with these values.[66] Table X.45 shows the projection results for soil management emissions.

66 County of San Diego: 2017 County of San Diego Crop Statistics and Annual Report, 2018 County of San Diego Crop Statistics and Annual Report, 2019 County of San Diego Crop Statistics and Annual Report. California Department of Food and Agriculture: 2017 Fertilizing Materials Tonnage Report, 2018 Fertilizing Materials Tonnage Report, 2019 Fertilizing Materials Tonnage Report, accessed on April 22, 2020.

Table X.45: Projected Greenhouse Gas Emissions from Soil Management

Projected Greenhouse Gas Emissions from Soil Management					
Projection Year	**2025**	**2030**	**2035**	**2045**	**2050**
Oats Harvested (acres)	2,091	2,131	2,172	2,255	2,298
Crop Residue Nitrogen (kg N)	7,996	8,176	8,359	8,738	8,935
Farm Nitrogen (kg N)	2,545,395	2,639,806	2,737,717	2,944,571	3,053,786
Non-Farm Nitrogen (kg N)	3,034,402	3,148,316	3,266,506	3,516,362	3,648,369
N_2O Emitted per Unit of Nitrogen (kg N_2O-N/kg N)	0.01	0.01	0.01	0.01	0.01
Farm Nitrogen Volatilized (kg N)	254,540	263,981	273,772	294,457	305,379
Non-Farm Nitrogen Volatilized (kg N)	303,440	314,832	326,651	351,636	364,837
N_2O Emitted per Unit of Nitrogen Volatilized (kg N-N_2O/kg NH_3- N + NOx-N volatilized)	0.01	0.01	0.01	0.01	0.01
Farm Nitrogen Leached (kg N)	763,619	791,942	821,315	883,371	916,136
Non-Farm Nitrogen Leached (kg N)	910,321	944,495	979,952	1,054,909	1,094,511
N_2O Emitted per Unit of Nitrogen Leached/Runoff (kg N_2O-N/kg N leaching/runoff)	0.0075	0.0075	0.0075	0.0075	0.0075
Amount on Lime Applied (tons)	195	198	200	206	208
Carbon Content of Lime (ton C/ton of lime)	0.125	0.125	0.125	0.125	0.125
Amount of Urea Applied (tons)	500	508	516	532	540
Carbon Content of Urea (ton C/ton of urea)	0.2	0.2	0.2	0.2	0.2
Total Farm Emissions (MMT CO_2e)	0.02	0.02	0.02	0.02	0.02
Total Non-Farm Emissions (MMT CO_2e)	0.02	0.02	0.02	0.02	0.02
Total GHG Emissions (MMT CO_2e)	**0.04**	**0.04**	**0.04**	**0.04**	**0.04**

Source: EPIC, USD 2020.