

Methodology For Illinois Electric Customers and Sales Forecasts: 2017-2026

In December 2014, an electric rate case was finalized in MEC's Illinois service territory. As a result of the implementation of new electric rates, a number of customers were switched to a different revenue class. This switching will cause noticeable changes in the forecast, as compared to historical values.

The 2017-2026 electric customer and sales forecasts were produced using econometric models on a monthly basis and are carried out in three steps using a top-down approach:

Step 1: The aggregate customer numbers were forecasted directly by revenue class:

- Residential
- Commercial
- Industrial
- Public authority.

Industrial kWh sales were forecast directly. The street lighting forecasts were forecast using trending. In this class, the current customer numbers were assumed to remain constant while the corresponding energy sales were projected to grow approximately 0.20% annually in IL. Similar to the peak demand forecast, the Quad Cities' economic and demographic drivers are assumed to be a good proxy for MidAmerican Illinois service territory electric sales and customers in these forecasts.

Step 2: For residential, commercial and public authority, econometric models were built to forecast kWh per customer. The resulting kWh per customer forecasts were multiplied by the appropriate customer forecasts to arrive at a kWh sales forecast. For industrial, the kWh per customer values for each revenue class were calculated using customer and sales forecasts, and employed to check the presence of any discontinuity between the historical and forecasted values.

Step 3: The projected customers and sales numbers were modeled using data specific to the area being forecast. Economic data for the Quad Cities' metropolitan statistical area was used in building the models.

Economic and demographic variables

Some variables, such as customer numbers, price, sales, revenue class, jurisdiction, etc., were obtained internally from the company database while other data, such as economic, demographic and weather, were received from external sources.

The economic and demographic data for the models were obtained from the IHS Global Insight, Inc. database. The economic and demographic data forecast was performed by IHS Global Insight, Inc. in March 2016. The list of variables considered for the electric sales and customer forecasts is shown in Table 1. For MEC's Illinois service territory, economic and demographic variables specific to the Quad Cities metropolitan area were used in the forecasting process. The Quad Cities area encompasses MEC's Illinois service territory.

Table 1: List of economic and demographic variables considered for the 2017-2026 forecasts

Quad Cities MSA	
1	Real Gross Metropolitan Area Product (Millions 2009\$)
2	Real Gross Metropolitan Area Product, Government, State and Local (Millions 2009\$)
3	Real Gross Metropolitan Area Product, Manufacturing (Millions 2009\$)
4	Population (Thousands)
5	Households, Family and Non-Family (Thousands)
6	Employment (NAICS), Total Non-Farm (Thousands)
7	Employment (NAICS), State and Local Government (Thousands)

Weather variables

The weather variables (derived from conditions at the Moline International Airport) used in the present forecast are:

- Current month and previous month cooling degree days (CDD)
- Current month and previous month heating degree days (HDD)

The present energy forecasts are based on billed data. This means that the sales numbers reflect, in part, the weather conditions from the previous month as well as the weather conditions for the current month, depending on the meter read date. To take this into account, both current month and previous month degree days are used in the modeling process. The forecasts used actual weather values for the historical period and normal weather values for the forecast period. In the 2017-2026 forecast, normal weather was defined as the average monthly degree days from 1986-2015.

To compare the growth rates the historical sales figures were “weather normalized” using average (normal) weather values. The normalization process consists of three steps. First, the historic predicted numbers were obtained from a regression model using the actual weather values. Second, the sales were re-calculated using average weather results.¹ Third, the difference between them, which defines the weather impact, was subtracted from the corresponding actual sales to arrive the normalized sales. In mathematical terms, the weather normalization can be written as follows:

$$NormalizedSales = ActualSales - [PredictedSales_{ActualWeather} - PredictedSales_{NormalWeather}]$$

¹ The same equation obtained in the first step was used.

Modeling

The econometric forecasting method used in this study assumes that the relationship between the dependent and independent variables is linear (additive) and defined as follows:

$$y = r + \alpha X + \beta Y + \gamma Z$$

where X, Y and Z are the variables, α , β and γ are the coefficients and r is the constant.

The forecasts were prepared using MetrixND software, version 4.5.1, developed by Itron, Inc. The forecasts typically involve finding a mathematical relationship between the dependent and independent variables. The steps taken in this forecast were as follows: The historical numbers since 2000 and the forecast numbers for economic variables until 2045 were obtained. These values were then exported into MetrixND and the analysis was carried out.

The primary criterion in selecting the variables was the relevance to the dependent variable being forecasted. Other considerations were the sign (the direction of change) and impact (the magnitude of elasticity coefficients) of variables on the forecasted dependent variable. Some of the statistical parameters important to the econometric model are:

Adjusted R-Square: It indicates the fraction of total variation explained by the independent variables in the regression. Its value ranges between 0 and 1, 1 being a perfect fit.

$$R^2 = \frac{\text{Explained Variation}}{\text{Total Variation}}$$

Adjusted R^2 takes into account the number of variables (k) with a constant sample size (n) as this leads to a decrease in the degree of freedom (n-k). Thus, adjusted R^2 is more conservative.

$$\text{Adjusted } R^2 = 1 - (1 - R^2) \left(\frac{n-1}{n-k} \right)$$

F-Statistics (Probability): This is an alternative measure of goodness of the fit. F-statistics number indicates the probability that the estimated regression fit is purely accidental. This number is preferred to be as low as possible as compared to a critical number of 5%.

Mean Absolute Percentage Error (MAPE): MAPE defines the magnitude of errors in the model. It is the average of absolute values of the residual error percentages measured at each data point. The lower the MAPE number the better the model is considered to be.

Durbin-Watson Statistic: It tests the hypothesis that the errors from a model do not exhibit first order autocorrelation. In the absence of autocorrelation, the statistic has a value of 2. While it

varies between 0 and 4, a value above 2 indicates negative autocorrelation, while a value below 2 indicates positive autocorrelation.

Test parameters for statistical significance

The t-statistics and P-values show the statistical significance of independent variables in 95% confidence interval (or 5% significance level). Most of the variables presented in this document are within the 95% confidence interval based on the t-statistics and P-values.

To evaluate the reasonableness of the model, the residual patterns and model fit statistics were studied. The residuals indicate the difference between the predicted and actual values. Any pattern associated with residuals suggests a missing variable(s). The residuals were studied through the autocorrelation factor and partial autocorrelation diagrams.

Customer forecasts

Variables and model statistics

The customer forecasts in general were straight-forward and involved fewer variables. The customer variables used in the models of different revenue classes are:

- Residential: Number of households in the Quad Cities Metropolitan Statistical Area (MSA), binary variable for the Illinois rate case impact and monthly binary variables
- Commercial: Economic variable weighted between non-farm employment and real gross metropolitan area product in the Quad Cities MSA, binary variables for the Illinois rate case impact and monthly binary variables
- Industrials: Non-farm employment for the Quad Cities MSA, binary variable for the Illinois rate case impact and monthly binary variables
- Public authority: Economic variable weighted between state and local government employment in the Quad Cities MSA and non-farm employment in the Quad Cities MSA, binary variable for the Illinois rate case impact and monthly binary variables

The statistics for the customer forecasts are tabulated in Table 2.

Table 2: Adjusted R² and MAPE values for the customer forecasts

Revenue Class	MAPE
Residential	0.04%
Commercial	0.11%
Industrial	0.79%
Public Authority	0.41%

Customer forecast results

The monthly customer numbers are shown below at an average annual level for each revenue class.

Table 3: Summary of the historical and forecast average annual customer numbers in different classes

	Residential	Commercial	Industrial	Public Authority	Street Lighting	Total
2008	75,394	7,796	90	1,303	48	84,631
2009	75,497	7,774	91	1,333	48	84,743
2010	75,437	7,727	101	1,363	48	84,675
2011	75,516	7,721	104	1,427	44	84,813
2012	75,693	7,716	107	1,376	44	84,936
2013	75,765	7,709	105	1,389	44	85,012
2014	75,814	7,782	99	1,392	44	85,131
2015	74,455	8,998	56	1,302	42	84,852
2016	74,203	9,124	57	1,285	42	84,712
2017	74,274	9,102	58	1,287	42	84,763
2018	74,350	9,083	58	1,289	42	84,822
2019	74,417	9,069	59	1,291	42	84,877
2020	74,502	9,058	59	1,293	42	84,954
2021	74,608	9,050	60	1,294	42	85,054
2022	74,706	9,044	60	1,296	42	85,148
2023	74,829	9,040	60	1,297	42	85,268
2024	74,965	9,036	60	1,299	42	85,402
2025	75,106	9,033	61	1,300	42	85,541
2026	75,273	9,031	61	1,301	42	85,707

Sales forecasts

Variables and model statistics

The energy forecasts are more complicated and involve more variables than do the customer forecasts. For the residential, commercial and public authority classes, sales are determined by multiplying customers by use per customer. For the industrial class, sales are modeled directly. For the street lighting class, sales are forecast using trending. The sales forecast variables used in the industrial class model are:

- Industrial: An weighted index made up of the real gross metropolitan area product for the Quad Cities MSA, the non-farm employment in the Quad Cities MSA and the population of the Quad Cities MSA, current month cooling degree days, industrial retail average revenue lagged twelve months, and monthly binaries.

The statistics for the sales forecasts are tabulated in Table 4.

Table 4: Adjusted R^2 and MAPE values for the sales forecasts

Revenue Class	MAPE
Industrial	3.33%

The comparison of tables (Tables 2 and 4) clearly indicates that better statistics were obtained for the customer models than sales models. The reason is that there is more uncertainty in the sales forecasts due to the presence of multiple drivers and their possible interactions. For example, a relatively small change in the historical usage pattern of a large industrial customer could have a measureable impact on the total energy usage in this class. Similarly, the changes in billing cycle could have significant effect on the billed sales.

Sales forecast results

The monthly billed sales numbers were forecasted at an aggregate level for each revenue class. The annual historical data and 10-year forecast values are summarized in Table 5.

Table 5: Summary of the historical and forecast annual billed sales of different revenue classes (MWh)

	Residential	Commercial	Industrial	Public Authority	Street Lighting	Total
2008	634,396	444,459	679,629	192,841	13,332	1,964,657
2009	642,453	438,115	643,796	191,417	13,257	1,929,038
2010	664,574	439,390	691,456	201,216	13,319	2,009,955
2011	661,451	436,720	714,016	203,850	12,911	2,028,948
2012	668,265	438,307	712,702	191,446	12,647	2,023,366
2013	665,762	435,113	686,082	185,062	12,599	1,984,618
2014	665,362	430,923	681,658	177,018	12,595	1,967,556
2015	627,826	461,907	641,935	163,747	10,129	1,905,544
2016	638,388	459,221	641,516	165,395	10,449	1,914,969
2017	625,949	454,836	653,346	164,376	10,483	1,908,991
2018	625,035	451,070	657,589	164,912	10,485	1,909,091
2019	622,917	447,733	660,243	165,472	10,465	1,906,830
2020	619,927	444,010	665,418	165,976	10,487	1,905,818
2021	616,631	440,653	671,537	166,582	10,516	1,905,919
2022	613,761	436,668	677,597	167,108	10,540	1,905,674
2023	610,676	432,907	682,583	167,731	10,572	1,904,469
2024	607,677	429,382	688,179	168,346	10,604	1,904,190
2025	605,620	425,947	693,235	168,850	10,634	1,904,287
2026	603,797	422,582	698,309	169,173	10,654	1,904,515
The figures in the table above are retail billed MWh sales.						

Usage per customer (UPC) forecasts

For the residential, commercial and public authority classes, kWh per customer values was forecast using econometric models. For the industrial and street lighting classes, the kWh per customer forecast values were calculated using the forecast sales and customer numbers data.

UPC forecast results:

Residential model – Economic variable (weighted between members per household in the Quad Cities metropolitan area and real per capita income in the Quad Cities MSA), billing days, cooling degree days (current month), heating degree days (current month), residential average revenue, binary variable for the Illinois rate case impact and monthly binaries

Commercial model – Number of households for the Quad Cities MSA, cooling degree days (current month), heating degree days (current and lagged month), a trend variable multiplied by the current month heating degree days, hours of light, commercial average revenue, billing days, hours of light, binary variable for the Illinois rate case impact and monthly binaries

Public Authority model – State and local government employment, billing days, heating degree days (current month), cooling degree days (current month), hours of light, binary variable for the Illinois rate case impact and monthly binaries

Table 6: Model Statistics

Revenue Class	MAPE
Residential	2.77%
Commercial	2.65%
Public Authority	5.04%

Methodology for the Monthly Illinois Non-Coincident Electric Gross Peak Demand Forecast: 2017-2026

2015 Electric Gross Peak Demand

The gross peak numbers used in the analysis are the historical gross peaks, which take into account demand side management impacts. Since there are planned large load additions, using the model results alone for the peak demand forecast would result in a forecast that is too low. Therefore, the planned large load additions are added to the model results to achieve the final peak demand forecast.

The gross peak load value was calculated according to the following equation:

$$\text{Gross Peak} = \text{Native Peak Load} + \text{Residential Direct Load Control} + \text{Curtailment}$$

Native Peak Load: For MEC's Illinois service territory, the 2015 native system peak load of 437 MW occurred on July 13, 2015 in the hour ending at 5:00 p.m. Central Daylight Time.

SummerSaver Program: SummerSaver is MEC's residential direct load control program. Load displaced due to the energy saving program which aims to curtail energy usage of on-peak hours was also received from the energy efficiency group. At the time of gross system peak, the SummerSaver program was not in effect.

Curtailment: Load displaced due to curtailment of customers on an interruptible rate. There was no curtailment event in effect at the time of gross system peak.

Source Data and Model

The historical hourly data underlying the model is load research data by class for MEC's Illinois service territory. The data was divided into the following classes: residential, small commercial, large commercial, small industrial and large industrial. This data was at the meter level. MEC used data from January 1, 2008 through December 31, 2015 to build a monthly non-coincident electric gross peak demand model for its Illinois service territory.

The class data was added together to derive the total Illinois load. Next, the monthly peak dates and times were calculated. Weather data, taken from the weather station at the Quad City International Airport in Moline, IL, associated with the peak dates were compiled for use in the model.

The forecasting model consists of an economic driver variable, a number of weather variables and monthly indicator variables.

Economic variables

Net Energy for Load multiplied by a Time Trend Variable

For the 2017-2026 forecast, MEC used the area's net energy for load multiplied by a time trend variable as the economic driver. This variable was constructed in the following manner:

Net Energy for Load*Time Trend

Weather variables

Five weather variables were used:

1. Summer peak day maximum temperature (summer = May through September)
2. Summer peak day average daily dew point
3. Winter peak day minimum temperature (winter = November through March)
4. Winter peak day three day build up (the sum of the average temperatures of the three days prior to the winter peak day)
5. Shoulder peak day HDD65 (shoulder = April and October; HDD65 = 65 less the peak day average temperature, if the average temperature is less than 65; = 0 if the average temperature is greater than 65)

The forecast weather was calculated using the rank and average method for 2008 through 2015. First, the weather variables, as measured on the monthly peak days, were averaged for each month across the years. This revealed the monthly order for each weather variable throughout the year. For each year, the peak day weather variables were then ranked. Next, the ranked results were averaged: the highest values averaged, the second highest values averaged, and so on. The average of the highest values was then assigned to the month with the highest value, the average of the second highest values was then assigned to the month with the second highest value and so on.

Table 7: MEC Illinois monthly non-coincident peak demand forecast

Year	Month	Peak
2016	1	315.403
2016	2	290.722
2016	3	275.091
2016	4	264.500
2016	5	316.212
2016	6	410.305
2016	7	451.390
2016	8	440.376
2016	9	386.248
2016	10	276.230
2016	11	278.571
2016	12	305.391
2017	1	316.399
2017	2	290.639
2017	3	275.374
2017	4	264.191
2017	5	316.285
2017	6	411.325
2017	7	453.865
2017	8	442.551
2017	9	386.850
2017	10	276.261
2017	11	278.532
2017	12	306.097
2018	1	317.562
2018	2	290.611
2018	3	275.654
2018	4	263.789
2018	5	316.224
2018	6	412.325
2018	7	456.519
2018	8	444.830
2018	9	387.354
2018	10	276.232
2018	11	278.515
2018	12	306.964
2019	1	318.637
2019	2	290.518
2019	3	275.859
2019	4	263.325
2019	5	316.111
2019	6	413.272
2019	7	459.093
2019	8	447.040
2019	9	387.817
2019	10	276.130
2019	11	278.419
2019	12	307.734
2020	1	319.755
2020	2	290.475
2020	3	276.142
2020	4	262.953
2020	5	316.093
2020	6	414.285
2020	7	461.705
2020	8	449.292
2020	9	388.347
2020	10	276.120
2020	11	278.399
2020	12	308.560
2021	1	320.916
2021	2	290.478
2021	3	276.490
2021	4	262.652
2021	5	316.144
2021	6	415.346
2021	7	464.355
2021	8	451.588
2021	9	388.932
2021	10	276.183
2021	11	278.440
2021	12	309.438
2022	1	322.058
2022	2	290.462
2022	3	276.819
2022	4	262.336
2022	5	316.186
2022	6	416.403
2022	7	466.992
2022	8	453.864
2022	9	389.499
2022	10	276.229
2022	11	278.467
2022	12	310.299

Weather in the Hourly Model

Using average daily temperature as an example, this is how a chaotic normal weather pattern (weather pattern used to create a realistic 8760 for dispatch simulations) is created:

1. Sort the Order variable (a ranking of the days in the month by average temperature, determined over the 1986-2015 time period) and the associated dates from highest to lowest within each month.
2. Sort the average temperature variable from highest to lowest within each month.
3. Assign the highest average temperature value to the date that corresponds to the highest value in the Order variable within the month.
4. Sort the Order variable by date for each month.
5. Create the average temperature output variable for the reference year.
6. Rotate the average temperature output variable to multiple years for forecasting purposes.

Hourly Loadshape Models by Class

Hourly models by class (residential, commercial, industrial, public authority and street lighting) were developed in MetrixND. The source data was hourly load research data by class for MEC's service territory. The classes of load research data were residential, small commercial, large commercial, small industrial and large industrial. The residential class load shape was developed using the residential load research data. The commercial class loadshape was developed by combining the small and large commercial load research data. The industrial class loadshape was developed using the small and large industrial load research data. The street lighting loadshape was a lighting loadshape from MEC's load research library. The public authority class loadshape was developed by using a weighted average of the residential, commercial, industrial and street lighting class loadshapes, based on the rate codes that made up the public authority class. Making use of linear regression, the models were estimated on data from January 1, 2012 through December 31, 2015. The models contain weather, binary and trend explanatory variables. There were twenty four models for each class. A forecast was developed through December 31, 2022, using the weather forecast developed as described above.

Long-Term Hourly Modeling

The long-term hourly forecast was developed in MetrixLT. The hourly profiles by class were calibrated to existing calendar month sales forecasts by class and an overall monthly non-coincident peak demand forecast.

Energy Efficiency in the Load Forecast

MEC has energy efficiency programs operating in its Illinois service territory. Estimated past energy savings are implicit in the historical data used to derive the electric sales forecast models. Without adjustment, this method implies that the level of future estimated program savings will be similar to past estimated program savings. Estimated program impacts in the forecast period are not projected to deviate measurably from estimated historical levels, so no adjustment was made to the forecasting models.

Load Forecast for the Retail Choice Switching

MEC has one active alternative retail supplier in its Illinois service territory. The retail choice switching forecast was derived by reviewing recent switching activity and projecting forward recent trends. Switched load is expected to grow from 11.5 MW in 2016 to 11.7 MW in 2022.

Table 8: Retail Switching: Monthly Peak Demand and Energy Forecasts

	Residential kWh	Commercial kWh	Industrial kWh	Public Authority kWh	Street Lighting kWh	Total kWh	MW Demand
Jan-16	47,946	3,069,954	1,205,865	2,867	240,688	4,567,320	8.51
Feb-16	50,878	3,050,985	1,056,305	2,411	224,113	4,384,692	8.50
Mar-16	47,754	2,841,158	964,111	2,181	188,346	4,043,550	8.68
Apr-16	33,914	3,030,588	1,095,655	2,452	167,852	4,330,460	8.28
May-16	31,108	2,902,453	1,100,927	2,526	139,053	4,176,068	9.20
Jun-16	41,045	3,370,497	1,110,956	2,764	99,142	4,624,404	10.02
Jul-16	55,727	3,877,830	1,151,534	2,880	76,301	5,164,273	11.50
Aug-16	58,618	3,952,104	1,156,894	3,083	89,394	5,260,093	11.19
Sep-16	45,537	3,544,781	1,102,226	2,873	132,411	4,827,828	10.09
Oct-16	35,904	3,383,387	1,047,227	2,554	188,847	4,657,918	8.51
Nov-16	32,438	2,714,547	1,032,953	2,110	192,295	3,974,343	8.63
Dec-16	44,421	3,194,743	1,018,506	2,426	240,348	4,500,444	9.20
Jan-17	48,018	3,052,411	1,209,484	2,870	240,808	4,553,592	8.54
Feb-17	50,954	3,091,150	1,059,475	2,413	224,225	4,428,218	8.53
Mar-17	47,826	2,849,685	967,005	2,183	188,440	4,055,139	8.71
Apr-17	33,964	3,039,683	1,098,943	2,455	167,936	4,342,981	8.30
May-17	31,154	2,911,164	1,104,232	2,529	139,123	4,188,202	9.23
Jun-17	41,107	3,380,612	1,114,291	2,767	99,192	4,637,968	10.05
Jul-17	55,811	3,889,469	1,154,990	2,883	76,340	5,179,492	11.53
Aug-17	58,706	3,963,965	1,160,366	3,086	89,439	5,275,562	11.22
Sep-17	45,605	3,555,420	1,105,534	2,876	132,477	4,841,912	10.12
Oct-17	35,958	3,393,541	1,050,370	2,556	188,941	4,671,366	8.54
Nov-17	32,487	2,722,694	1,036,053	2,112	192,391	3,985,737	8.65
Dec-17	44,487	3,204,332	1,021,562	2,429	240,468	4,513,278	9.23
Jan-18	48,090	3,061,572	1,213,114	2,873	240,929	4,566,578	8.57
Feb-18	51,031	3,100,427	1,062,655	2,416	224,337	4,440,866	8.55
Mar-18	47,897	2,858,238	969,907	2,185	188,534	4,066,762	8.73
Apr-18	34,015	3,048,806	1,102,241	2,457	168,020	4,355,540	8.33
May-18	31,201	2,919,901	1,107,546	2,531	139,192	4,200,372	9.26
Jun-18	41,168	3,390,759	1,117,635	2,770	99,241	4,651,573	10.08
Jul-18	55,894	3,901,142	1,158,456	2,886	76,378	5,194,756	11.57
Aug-18	58,794	3,975,862	1,163,849	3,089	89,484	5,291,078	11.25
Sep-18	45,674	3,566,091	1,108,852	2,879	132,543	4,856,039	10.15
Oct-18	36,012	3,403,726	1,053,522	2,559	189,036	4,684,855	8.56
Nov-18	32,536	2,730,865	1,039,163	2,115	192,487	3,997,165	8.68
Dec-18	44,554	3,213,949	1,024,628	2,431	240,588	4,526,151	9.26
Jan-19	48,162	3,070,761	1,216,755	2,876	241,049	4,579,603	8.59
Feb-19	51,107	3,109,732	1,065,844	2,418	224,449	4,453,552	8.58
Mar-19	47,969	2,866,816	972,818	2,188	188,629	4,078,419	8.76
Apr-19	34,066	3,057,957	1,105,549	2,460	168,104	4,368,136	8.35
May-19	31,248	2,928,665	1,110,870	2,534	139,262	4,212,578	9.28
Jun-19	41,230	3,400,935	1,120,989	2,772	99,291	4,665,218	10.11
Jul-19	55,978	3,912,850	1,161,933	2,889	76,416	5,210,066	11.60
Aug-19	58,883	3,987,795	1,167,342	3,092	89,528	5,306,639	11.29
Sep-19	45,742	3,576,793	1,112,180	2,882	132,609	4,870,207	10.18
Oct-19	36,066	3,413,941	1,056,684	2,561	189,130	4,698,383	8.59
Nov-19	32,585	2,739,061	1,042,282	2,117	192,583	4,008,627	8.70
Dec-19	44,621	3,223,594	1,027,704	2,434	240,709	4,539,061	9.28
Jan-20	48,234	3,079,977	1,220,407	2,878	241,170	4,592,666	8.62
Feb-20	51,184	3,119,065	1,069,043	2,421	224,562	4,466,275	8.61
Mar-20	48,041	2,875,420	975,737	2,190	188,723	4,090,111	8.78
Apr-20	34,118	3,067,134	1,108,867	2,462	168,188	4,380,769	8.38
May-20	31,295	2,937,454	1,114,204	2,536	139,332	4,224,821	9.31
Jun-20	41,292	3,411,142	1,124,354	2,775	99,341	4,678,903	10.14
Jul-20	56,062	3,924,594	1,165,420	2,892	76,454	5,225,422	11.64
Aug-20	58,971	3,999,763	1,170,845	3,095	89,573	5,322,247	11.32
Sep-20	45,811	3,587,528	1,115,518	2,885	132,676	4,884,417	10.21
Oct-20	36,120	3,424,188	1,059,855	2,564	189,225	4,711,952	8.61
Nov-20	32,633	2,747,282	1,045,410	2,119	192,680	4,020,123	8.73
Dec-20	44,688	3,233,269	1,030,788	2,436	240,829	4,552,010	9.31
Jan-21	48,307	3,089,221	1,224,069	2,881	241,290	4,605,769	8.64
Feb-21	51,261	3,128,427	1,072,252	2,423	224,674	4,479,036	8.63
Mar-21	48,113	2,884,050	978,666	2,192	188,817	4,101,838	8.81
Apr-21	34,169	3,076,339	1,112,195	2,464	168,272	4,393,440	8.40
May-21	31,342	2,946,271	1,117,548	2,539	139,401	4,237,100	9.34
Jun-21	41,354	3,421,380	1,127,728	2,778	99,390	4,692,630	10.17
Jul-21	56,146	3,936,372	1,168,918	2,895	76,492	5,240,824	11.67
Aug-21	59,060	4,011,767	1,174,359	3,098	89,618	5,337,902	11.36
Sep-21	45,880	3,598,295	1,118,866	2,887	132,742	4,898,670	10.25
Oct-21	36,174	3,434,464	1,063,036	2,567	189,319	4,725,561	8.64
Nov-21	32,682	2,755,527	1,048,547	2,121	192,776	4,031,654	8.76
Dec-21	44,755	3,242,973	1,033,882	2,439	240,949	4,564,998	9.34
Jan-22	48,379	3,098,492	1,227,743	2,884	241,411	4,618,910	8.67
Feb-22	51,338	3,137,816	1,075,470	2,426	224,786	4,491,835	8.66
Mar-22	48,186	2,892,706	981,603	2,194	188,912	4,113,600	8.84
Apr-22	34,220	3,085,572	1,115,533	2,467	168,356	4,406,149	8.43
May-22	31,389	2,955,113	1,120,902	2,541	139,471	4,249,416	9.37
Jun-22	41,416	3,431,648	1,131,113	2,781	99,440	4,706,397	10.20
Jul-22	56,231	3,948,186	1,172,426	2,898	76,531	5,256,271	11.71
Aug-22	59,148	4,023,807	1,177,884	3,101	89,663	5,353,603	11.39
Sep-22	45,948	3,609,095	1,122,224	2,890	132,808	4,912,966	10.28
Oct-22	36,228	3,444,772	1,066,227	2,569	189,414	4,739,210	8.66
Nov-22	32,731	2,763,797	1,051,694	2,123	192,872	4,043,218	8.78
Dec-22	44,822	3,252,706	1,036,985	2,441	241,070	4,578,024	9.37

Table 9: Retail Switching: Monthly Customer Count Forecasts

	Residential	Commercial	Industrial	Public Authority	Street Lighting	Total
Jan-16	97	322	3	8	2	432
Feb-16	96	316	3	8	2	425
Mar-16	96	320	3	8	2	429
Apr-16	97	320	3	8	2	430
May-16	97	320	3	8	2	430
Jun-16	97	320	3	8	2	430
Jul-16	98	320	3	8	2	431
Aug-16	97	320	3	8	2	430
Sep-16	96	320	3	8	2	429
Oct-16	96	320	3	8	2	429
Nov-16	97	320	3	8	2	430
Dec-16	97	320	3	8	2	430
Jan-17	97	320	3	8	2	430
Feb-17	96	320	3	8	2	429
Mar-17	96	321	3	8	2	430
Apr-17	97	321	3	8	2	431
May-17	97	321	3	8	2	431
Jun-17	97	321	3	8	2	431
Jul-17	98	321	3	8	2	432
Aug-17	97	321	3	8	2	431
Sep-17	96	321	3	8	2	430
Oct-17	96	321	3	8	2	430
Nov-17	97	321	3	8	2	431
Dec-17	97	321	3	8	2	431
Jan-18	97	321	3	8	2	431
Feb-18	96	321	3	8	2	430
Mar-18	96	322	3	8	2	431
Apr-18	97	322	3	8	2	432
May-18	97	322	3	8	2	432
Jun-18	97	322	3	8	2	432
Jul-18	98	322	3	8	2	433
Aug-18	97	322	3	8	2	432
Sep-18	96	322	3	8	2	431
Oct-18	96	322	3	8	2	431
Nov-18	97	322	3	8	2	432
Dec-18	97	322	3	8	2	432
Jan-19	97	322	3	8	2	432
Feb-19	96	322	3	8	2	431
Mar-19	96	322	3	8	2	432
Apr-19	97	322	3	8	2	433
May-19	97	322	3	8	2	433
Jun-19	97	322	3	8	2	433
Jul-19	98	322	3	8	2	434
Aug-19	97	322	3	8	2	433
Sep-19	96	322	3	8	2	432
Oct-19	96	322	3	8	2	432
Nov-19	97	322	3	8	2	433
Dec-19	97	322	3	8	2	433
Jan-20	97	322	3	8	2	433
Feb-20	96	322	3	8	2	432
Mar-20	96	323	3	8	2	433
Apr-20	97	323	3	8	2	434
May-20	97	323	3	8	2	434
Jun-20	97	323	3	8	2	434
Jul-20	98	323	3	8	2	435
Aug-20	97	323	3	8	2	434
Sep-20	96	323	3	8	2	433
Oct-20	96	323	3	8	2	433
Nov-20	97	323	3	8	2	434
Dec-20	97	323	3	8	2	434
Jan-21	97	323	3	8	2	434
Feb-21	96	323	3	8	2	433
Mar-21	96	324	3	8	2	434
Apr-21	98	324	3	8	2	435
May-21	98	324	3	8	2	435
Jun-21	98	324	3	8	2	435
Jul-21	99	324	3	8	2	436
Aug-21	98	324	3	8	2	435
Sep-21	97	324	3	8	2	434
Oct-21	97	324	3	8	2	434
Nov-21	98	324	3	8	2	435
Dec-21	98	324	3	8	2	435
Jan-22	98	324	3	8	2	435
Feb-22	97	324	3	8	2	434
Mar-22	97	325	3	8	2	434
Apr-22	98	325	3	8	2	436
May-22	98	325	3	8	2	436
Jun-22	98	325	3	8	2	436
Jul-22	99	325	3	8	2	437
Aug-22	98	325	3	8	2	436
Sep-22	97	325	3	8	2	435
Oct-22	97	325	3	8	2	435
Nov-22	98	325	3	8	2	436
Dec-22	98	325	3	8	2	436

Table 10: Multi-Year Historical Load Detail

	Small Industrial		Residential		Large Commercial		Small Commercial		Large Industrial		Lighting		Total	
	kWh	kW Demand	kWh	kW Demand	kWh	kW Demand	kWh	kW Demand	kWh	kW Demand	kWh	kW Demand	kWh	kW Demand
Jan-11	33,587,021	60,356	63,558,584	124,371	32,858,995	56,840	12,484,292	27,735	33,879,249	57,029			176,368,143	289,856
Feb-11	31,445,510	60,714	52,030,571	126,883	28,822,745	57,878	11,076,756	27,440	31,139,566	58,186			154,515,148	291,431
Mar-11	34,844,777	60,334	49,152,078	105,401	30,354,610	53,584	11,355,730	26,701	34,278,541	57,323			159,985,735	267,011
Apr-11	32,610,537	60,588	41,640,708	96,738	28,470,191	53,247	9,880,182	25,309	34,174,198	58,549			146,775,816	250,181
May-11	35,113,848	69,582	46,862,781	160,491	31,122,614	69,536	10,383,130	30,770	35,735,431	60,008			159,217,805	364,995
Jun-11	37,474,697	72,438	61,707,542	210,173	34,258,347	74,936	11,748,507	33,844	33,709,916	58,939			178,899,009	418,441
Jul-11	39,881,255	73,962	97,050,843	239,760	40,471,586	80,816	14,686,219	37,551	35,443,610	61,491			227,533,512	463,907
Aug-11	40,135,259	75,518	74,286,910	231,505	38,623,277	79,935	13,971,404	37,706	37,675,522	61,262			204,692,372	451,016
Sep-11	33,238,277	73,560	47,674,452	213,450	31,112,096	76,274	11,030,661	33,909	29,323,684	58,009			152,379,169	427,155
Oct-11	32,045,488	60,561	44,080,574	100,955	30,441,798	61,991	12,039,433	27,931	36,316,260	56,773			154,923,553	268,812
Nov-11	30,558,435	59,306	47,139,843	107,664	28,932,103	53,415	10,427,821	26,879	32,823,272	54,774			149,881,474	263,609
Dec-11	30,248,661	57,049	56,930,876	114,861	30,782,254	55,425	11,688,773	27,331	34,558,204	55,704			164,208,767	273,734
Jan-12	27,265,619	49,536	56,326,628	118,087	31,768,168	57,164	11,931,584	27,961	35,071,099	57,853			162,363,099	279,133
Feb-12	25,807,835	48,001	49,912,934	108,123	29,950,714	55,848	10,888,440	27,254	32,369,737	55,528			148,929,661	261,495
Mar-12	28,004,862	50,959	43,561,072	100,119	31,067,096	58,370	10,257,627	25,938	33,738,961	56,232			146,629,619	245,178
Apr-12	25,852,800	52,904	39,904,905	103,648	28,924,482	59,010	9,616,023	24,672	39,947,919	66,878			144,246,129	266,199
May-12	28,711,036	54,861	52,154,588	177,777	32,966,930	64,207	10,917,565	28,372	41,198,083	69,253			165,948,202	326,474
Jun-12	29,129,914	58,369	71,496,282	214,118	34,916,155	75,781	11,922,001	35,245	37,189,526	67,518			184,653,879	399,777
Jul-12	31,095,788	57,696	103,006,717	243,087	40,823,126	78,473	14,923,361	36,093	36,865,588	58,977			226,714,579	447,545
Aug-12	29,809,952	56,996	70,528,438	212,327	37,080,494	74,287	13,227,171	36,328	28,700,133	57,065			179,346,189	410,147
Sep-12	26,034,008	57,552	48,351,307	201,253	31,312,000	75,386	12,226,491	37,025	33,618,677	56,722			151,542,483	403,705
Oct-12	26,579,723	53,023	42,833,452	90,444	30,319,509	57,773	11,166,817	27,750	34,929,027	58,846			145,828,529	252,700
Nov-12	24,546,296	46,399	46,626,192	113,914	28,824,052	53,455	10,682,298	27,595	35,868,680	62,130			146,547,518	270,614
Dec-12	24,215,212	45,799	57,157,810	124,168	30,201,155	54,224	11,165,576	26,718	35,344,824	61,322			158,084,576	281,537
Jan-13	27,106,687	48,212	57,763,721	124,593	32,578,799	58,191	12,537,949	28,309	37,185,801	62,758			167,172,957	286,899
Feb-13	24,688,194	47,337	50,829,345	114,607	29,072,834	57,810	11,073,990	29,779	33,299,724	62,913			148,964,087	282,749
Mar-13	26,220,003	48,722	52,219,135	106,153	30,848,837	55,793	11,328,641	26,392	36,819,306	64,886			157,435,923	263,666
Apr-13	27,352,493	54,403	45,390,111	96,600	30,305,138	60,356	10,797,868	25,534	34,620,740	65,819			143,125,535	270,360
May-13	27,976,866	55,049	48,796,017	140,837	32,014,005	66,279	10,789,410	27,032	41,520,639	68,153			161,096,938	318,503
Jun-13	29,178,358	58,961	62,501,897	184,372	33,569,066	71,614	11,631,006	32,626	40,972,297	69,950			177,852,624	386,962
Jul-13	31,117,029	59,693	78,822,779	224,668	37,662,758	78,155	14,095,610	37,186	39,814,333	67,279			201,512,509	436,092
Aug-13	31,414,517	61,036	78,082,784	227,917	38,475,007	80,461	13,638,301	37,881	34,962,043	62,248			196,572,654	444,241
Sep-13	29,040,366	60,154	58,446,603	219,971	33,903,038	78,166	11,860,091	35,628	35,488,202	64,578			168,738,301	432,925
Oct-13	27,394,695	54,323	44,269,216	110,032	31,188,327	64,473	12,294,198	30,325	34,883,662	64,474			150,030,099	291,762
Nov-13	25,482,946	47,586	49,123,124	108,863	29,739,595	54,731	12,363,551	30,086	33,439,182	61,136			150,148,398	272,247
Dec-13	25,698,844	48,122	62,485,047	126,836	32,650,338	58,737	13,047,739	30,739	36,368,544	63,762			170,250,512	285,865
Jan-14	28,171,730	49,761	67,104,249	142,274	33,701,323	57,619	13,840,010	30,258	35,536,920	58,882			178,354,231	313,262
Feb-14	26,197,668	49,481	59,481,578	126,833	29,869,294	56,943	11,983,787	29,046	33,939,937	61,217			161,472,263	289,748
Mar-14	27,752,047	49,916	54,363,062	131,121	30,800,211	55,959	12,109,374	28,836	37,830,750	67,578			162,855,445	285,736
Apr-14	26,526,608	50,315	43,469,025	101,202	28,915,683	54,235	10,919,129	29,247	39,219,975	68,305			149,050,420	257,303
May-14	28,286,776	55,643	48,835,372	154,314	31,865,409	67,748	11,477,521	32,747	42,332,205	66,670			152,797,283	341,478
Jun-14	30,045,397	57,414	64,873,449	174,811	34,236,736	69,378	12,330,543	32,529	38,088,101	67,490			179,574,226	364,197
Jul-14	30,645,548	59,062	65,347,076	197,861	35,113,643	71,742	13,681,986	36,669	37,653,960	67,431			182,442,213	401,834
Aug-14	29,819,076	55,948	68,153,454	212,432	35,856,060	73,995	12,940,389	35,004	35,887,061	64,955			182,656,040	381,118
Sep-14	26,999,392	52,546	48,937,711	192,787	30,378,554	74,398	10,950,399	33,005	32,596,186	60,788			149,864,243	380,154
Oct-14	26,299,447	48,267	43,411,551	94,285	27,994,689	54,415	11,676,485	29,159	34,688,635	60,088			144,070,806	248,016
Nov-14	24,794,827	46,047	51,704,310	118,879	28,994,074	56,065	12,822,913	32,370	37,233,621	69,459			155,549,744	283,871
Dec-14	25,889,580	46,470	57,086,294	123,407	30,865,350	55,300	12,862,509	32,401	34,888,914	63,493			161,592,648	275,167
Jan-15	23,082,533	40,129	53,796,551	122,682	31,587,487	59,298	16,527,816	39,432	38,535,453	66,824	1,757,228	3,929	165,287,066	298,525
Feb-15	21,350,307	40,119	49,855,826	110,346	29,312,362	56,665	15,298,966	40,460	34,801,640	68,531	1,476,341	3,929	152,095,441	274,273
Mar-15	23,060,882	39,898	43,915,206	98,952	28,554,546	55,282	14,932,529	38,138	38,191,489	62,873	1,468,286	3,929	150,122,938	261,893
Apr-15	21,628,894	41,037	37,368,479	89,586	26,225,378	53,007	13,382,509	35,362	37,569,752	64,943	1,258,980	3,929	137,433,992	236,626
May-15	23,552,990	44,593	42,664,820	109,069	28,287,187	58,738	13,220,888	36,672	38,563,411	67,583	1,158,256	3,929	147,447,563	285,712
Jun-15	25,103,490	47,623	61,284,244	180,673	31,856,580	68,503	15,358,917	42,236	42,017,393	70,714	1,051,376	3,929	176,672,001	382,043
Jul-15	26,899,977	48,804	74,223,189	205,728	34,838,869	73,693	16,891,119	46,053	42,599,298	70,929	1,117,783	3,929	196,570,235	412,481
Aug-15	26,824,918	49,175	66,426,087	196,257	33,465,146	68,753	15,930,888	40,493	40,458,702	68,055	1,237,303	3,929	184,343,044	363,968
Sep-15	25,416,257	50,271	60,162,321	194,554	32,009,481	73,963	15,560,593	47,054	27,318,817	71,402	1,354,137	3,929	161,821,607	400,665
Oct-15	22,952,773	44,531	40,807,245	102,629	27,346,760	57,045	16,016,337	41,908	35,123,855	62,973	1,570,057	3,929	143,817,027	252,989
Nov-15	21,158,547	44,222	44,047,527	104,121	26,682,217	50,546	14,657,926	38,274	25,370,639	60,481	1,661,940	3,929	133,578,796	238,785
Dec-15	21,062,194	38,667	53,088,386	111,192	28,739,821	51,288	15,062,674	37,907	33,630,402	63,486	1,793,116	3,929	153,376,594	264,108

Low and High Load Forecast Scenarios

The required low and high hourly load forecast scenarios were created by taking the 95% confidence interval around each class-level sales, customer and use per customer forecast and the 95% confidence interval around the non-coincident gross peak demand forecast. MetrixND, the load forecasting software used for the sales, customers use per customer and non-coincident peak demand forecasts, provided the upper and lower bounds of a 95% confidence interval around each monthly forecast value. This software feature allowed the construction of upper and lower bound forecasts for the residential, commercial, industrial and public authority sales forecasts. The street lighting sales forecast was multiplied by 0.99 and 1.01 to generate, respectively, a lower and upper bound street lighting sales forecast. As mentioned above, the monthly residential, commercial and public authority sales forecasts were calculated by multiplying together a class-level customer forecast and a class-level use per customer forecast. For each month in the forecast period, the lower bound of each class-level sales forecast was found by multiplying the lower bound of the class-level customer count forecast by the lower bound of the class-level use per customer forecast. The same procedure was followed to arrive at the upper bound of the class-level sales forecasts. The industrial sales forecast was generated by a class-level total sales model. The lower and upper bounds of the 95% confidence interval were an output of the modeling process.

The lower bound forecasts of each class' 95% confidence interval were summed to arrive at the lower bound for the total sales forecast, while the upper bound forecasts of each class' 95% confidence interval were summed to arrive at the upper bound for the total sales forecast. The lower bound class-level sales forecasts were then applied to the appropriate load profile and, along with the lower bound non-coincident gross peak demand forecast, was run through MetrixLT to generate the lower bound of the hourly forecast. The same procedure was undertaken with the upper bound sales forecasts and non-coincident peak demand forecast to generate the upper bound of the hourly forecast.

The reference case temperature assumptions in the hourly load forecast model were not changed for the scenarios. The reference case weather-related assumptions in the sales, the use per customer and the non-coincident peak demand forecast models for MEC's Illinois service territory were not changed in the scenarios. The reference case forecasts for retail switching sales, customers and demand in MEC's Illinois service territory were not changed in the scenarios.