

**BEFORE THE STATE CORPORATION COMMISSION
OF THE STATE OF KANSAS**

**In the Matter of the Application of The)
Empire District Electric Company to)
Make Certain Changes in Its Charges) Docket No. 19-EPDE-223-RTS
for Electric Service)**

DIRECT TESTIMONY

PREPARED BY

DARREN L. PRINCE

UTILITIES DIVISION

KANSAS CORPORATION COMMISSION

May 13, 2019

1 **I. STATEMENT OF QUALIFICATIONS**

2 **Q. What is your name?**

3 A. Darren L. Prince

4 **Q. By whom and in what capacity are you employed?**

5 A. I am employed by the Kansas Corporation Commission (KCC or Commission) as
6 a Managing Economist.

7 **Q. What is your business address?**

8 A. 1500 SW Arrowhead Road, Topeka, KS, 66604-4027.

9 **Q. What is your educational background and professional experience?**

10 A. I hold a Bachelor of Arts and a Master of Arts degree in Economics from the
11 University of Missouri-Kansas City. I have been employed by the KCC since May
12 16, 2016. Prior to my employment by the KCC, I was a Project Manager for
13 Perfection Paper Hanging LLC.

14 **Q. Have you previously submitted testimony before this Commission?**

15 A. Yes, I filed testimony in Docket Nos. 16-KCPE-446-TAR, 16-KGSG-491-RTS,
16 18-WSEE-328-RTS, 18-KCPE-480-RTS, 18-KGSG-560-RTS, and 18-KG&E-
17 303-CON. I have also participated in other Dockets as a member of KCC Staff
18 (Staff).

19 **II. INTRODUCTION**

20 **Q. What is the purpose of your testimony?**

21 A. The purpose of my testimony is to sponsor Staff's weather normalization revenue
22 adjustment of \$(81,713) and Staff's customer annualization revenue adjustment of
23 \$40,417 for a total adjustment of \$(41,297).

1 **Q. Are you sponsoring any exhibits?**

2 A. Yes, I am sponsoring three exhibits. Exhibit DLP-1 is the regression analysis I
3 performed for each customer class, Exhibit DLP-2 is the weather normalization
4 adjustment, and Exhibit DLP-3 is the customer annualization adjustment.

5 **Q. How is your testimony organized?**

6 A. First, I will describe the purpose of normalizing revenue for weather. Then, I will
7 describe the process Staff utilizes to develop its weather normalization adjustment,
8 the results of Staff's analysis, and Staff's recommendation. Next, I will describe
9 the purpose of normalizing revenue for customer counts. Then, I will describe how
10 Staff developed its customer annualization adjustment, the results of Staff's
11 analysis, and Staff's recommendation. Finally, I will conclude by recommending
12 the Commission accept Staff's weather normalization adjustment and customer
13 annualization adjustment.

14 **III. ANALYSIS**

15 **WEATHER NORMALIZATION**

16 **Purpose of Weather Normalization**

17 **Q. What is the purpose of weather normalizing electricity usage?**

18 A. The weather normalization adjustment is designed to minimize the effects of non-
19 normal weather conditions on test year volumes and revenue collections. Some
20 electricity uses, such as air conditioning, are sensitive to temperature—increasing
21 when temperatures rise and decreasing when temperatures fall. Therefore, if the
22 test year is warmer than normal, volumes and revenue will be higher than normal.
23 Ultimately, this would result in rates being set too low for The Empire District

1 Electric Company (Empire) to collect its approved revenue requirement when
2 weather is normal.

3 Like air conditioning, electricity usage for space heating is also affected by
4 temperature. However, unlike air conditioning, which has a direct relationship with
5 temperature; space heating usage has an inverse relationship with temperature.
6 Thus, space heating usage increases when temperatures decrease and decreases
7 when temperatures increase.

8 Because test-year revenue should reflect normal ongoing operations, the
9 Commission sets rates based on normal weather customer usage. Through the
10 weather normalization process, test year volumes and revenues are adjusted to
11 reflect the difference between actual test year weather and normal weather. Hence,
12 weather normalization adjustments are added or subtracted to test year volumes and
13 revenue so the test year is reflective of normal weather.

14 **Process**

15 **Q. Please provide a brief description of the weather normalization process.**

16 A. The weather normalization process can be divided into the following four steps:

- 17 1) Selection of weather stations, the collection of weather data, and the collection
18 of customer usage data;
19 2) Performance of regression analysis to determine customer-class-specific
20 regression coefficients or weather sensitivity factors (WSFs);
21 3) Calculation of volumetric sales adjustments; and
22 4) Calculation of revenue adjustments.
23

1 Data Collection

2 **Q. Who provided the customer usage and customer count data?**

3 A. Empire provided Staff customer volumetric usage, customer count, and usage
4 charge revenue for each of the customer classes.¹ With this data, Staff was able to
5 calculate the per capita electric usage for each customer class and a monthly
6 weighted kilowatt hour (kWh) charge.

7 **Q. What is the source of weather data that Staff used for its analysis?**

8 A. Staff collected weather data for the Springfield, MO, weather station² from the
9 National Oceanic and Atmospheric Administration (NOAA). Staff collected daily
10 maximum and minimum temperature data for the period of January 1988 through
11 June 2018. Staff calculated Heating Degree Days (HDDs) and Cooling Degree
12 Days (CDDs) with this data. In addition, Staff calculated 30-year normals (1988-
13 2017).

14 **Q. Please explain what “HDDs” and “CDDs” are.**

15 A. HDDs and CDDs are weather variables that measure deviations from an established
16 base temperature (in this case, 65 degrees).³ HDDs cover colder weather and
17 indicate how cool the average daily temperature was relative to the base
18 temperature.⁴ In terms of weather normalization for electricity use, HDDs indicate

¹ KCC DR 87, KCC DR 87 Supplemental Response, KCC DR 87 Supplemental Updated, and KCC DR 133.

² The Springfield, MO weather station is the closest First Order weather station to Empire’s customers in Kansas.

³ Degree days are weather variables based on the assumption that when the outside temperature is 65 degrees Fahrenheit, an average person will not require heating or cooling to be comfortable.

https://www.weather.gov/key/climate_heat_cool

⁴ Staff calculated HDD and CDD measures as follows.

$$HDD = \left(65 - \frac{Max + Min}{2} \right) \text{ if } \frac{Max + Min}{2} < 65, \text{ otherwise } HDD = 0$$

1 customer demand for space heating. CDDs cover warmer weather and indicate how
2 warm the average daily temperature was relative to the base temperature. In terms
3 of weather normalization, CDDs indicate customer demand for air conditioning.

4 Regression Estimation Process

5 **Q. What is the purpose of performing a regression analysis on weather variables**
6 **and electricity use?**

7 A. The purpose of performing regression analysis is to determine the statistically
8 significant WSFs for each customer class. The WSFs measure the relationship
9 between customer usage (dependent variable) and weather (independent variables)
10 for the different customer classes. The WSFs are the estimated parameters in the
11 class regression equations for the HDD and CDD variables. The WSFs are then
12 used to calculate volume adjustments to correct for temperature deviations from the
13 30-year norms for each customer class.

14 **Q. What specifically were the customer usage and weather variables used in the**
15 **regression analysis.**

16 A. I performed a linear regression analysis with per capita customer usage as the
17 dependent variable. The monthly HDDs and lagged HDDs (HDD-1), CDDs and
18 lagged CDDs (CDD-1) were the independent variables. Because the data consists
19 of weather-sensitive variables collected at regular intervals over an extended period
20 of time, autocorrelation and seasonality was present in the data.⁵ Autocorrelation

$$CDD = \left(\frac{Max + Min}{2} - 65 \right) \text{ if } \frac{Max + Min}{2} > 65, \text{ otherwise } CDD = 0$$

⁵ Autocorrelation is the correlation of a time series variable with earlier and later value of itself. For example, the best predictor of next period US Gross Domestic Product (GDP) is this period's GDP plus or minus a small percentage change because US GDP is autocorrelated. Seasonality in time series data are regular patterns in the data. Air conditioning usage increases in the spring through the summer and then decreases in the fall through the winter.

1 and seasonality result in distortionary time series behavior; parameters such as the
2 mean and variance of the time series change over time.

3 **Q. How did you resolve the autocorrelation and seasonality issues?**

4 A. To correct for autocorrelation and seasonality, I applied autoregressive, seasonal
5 autoregressive, moving average, and seasonal moving average terms.⁶ Including
6 these terms substantially improved the standard error and other metrics of the data
7 analysis.

8 *Volumetric Adjustment Process*

9 **Q. Please describe the process for calculating the volumetric sales adjustments.**

10 A. To make the appropriate adjustment to the billing determinants, the actual monthly
11 HDDs and CDDs for the test year were subtracted from the normal monthly HDDs
12 and CDDs. The difference was multiplied by the WSFs and then finally multiplied
13 by the class customer count for a given month of the test year. The result is the
14 estimated effect on electricity usage, due to the deviation from normal weather.⁷

⁶ An AR(1) term is a first order autoregressive term and is defined as: $y_t = \rho_1 y_{t-1} + \varepsilon_t$. Notice that y_t is dependent upon the previous realization of y , y_{t-1} . The dependence y_t on y_{t-1} means that y_t is not only dependent on the current period error term but also on the previous period error term. The correlation of the error terms violates the classical regression assumption that the errors terms are uncorrelated. The use of the AR(1) term compensates for auto-correlated errors if the autocorrelation is of a first order autoregressive nature. An SAR(12) term is a seasonal autoregressive term with lag 12. It adds a polynomial with a lag of 12 to an existing AR specification. For example, an AR(1) and SAR(12) is defined as: $y_t = \rho_1 y_{t-1} + \varepsilon_t$ and $\phi_{12} y_{t-12} + \varepsilon_t$, and when combined, $y_t = \rho_1 y_{t-1} + \phi_{12} y_{t-12} - \rho_1 \phi_{12} y_{t-13} + \varepsilon_t$. The multiplication of the regular (ρ) and the seasonal (ϕ) autoregressive terms for the parameter y_{t-13} provides the non-linear effect.

An MA(1) term is a first order moving average term. It is defined as: $y_t = \varepsilon_t + \theta_1 \varepsilon_{t-1}$. Notice that y_t is dependent on both the current error term, ε_t , and the previous period error term, ε_{t-1} , thus the error terms are correlated. The MA(1) term compensates for the fact that the current period and previous period error terms are part of the equation. A SMA(12) term is a seasonal moving average term with lag 12. It adds a polynomial with lag of 12 to an existing MA specification. For example, an MA(1) and SMA(12) term is defined as: $y_t = \theta_1 \varepsilon_{t-1} + \varepsilon_t$ and $y_t = \omega_{12} \varepsilon_{t-12} + \varepsilon_t$, and when combined, $y_t = \theta_1 \varepsilon_{t-1} + \omega_{12} \varepsilon_{t-12} + \theta_1 \omega_{12} \varepsilon_{t-13} + \varepsilon_t$. The non-linearity is the same as in the autoregressive process except for the change in sign from minus to plus.

⁷ $\left(\frac{\text{Estimated}}{\text{effect on usage}} \right) = \left[\left(\left(\frac{\text{Normal}}{\text{HDD or CDD}} \right) - \left(\frac{\text{Actual}}{\text{HDD or CDD}} \right) \right) (WSF) \right] * (\text{Customer count})$

1 This calculation is done for each customer class, and the sum of all those
2 adjustments is the total weather normalized volumetric sales adjustment.

3 Revenue Adjustment Process

4 **Q. Please describe the process for calculating the revenue adjustment.**

5 A. To calculate the revenue adjustments, the volumetric sales adjustments for each
6 tariff class were multiplied by the appropriate rate for that customer class.⁸ The
7 result is the estimated revenue adjustment necessary to adjust test year revenues to
8 reflect weather normalized volumetric sales. The sum of all those adjustments is
9 the total weather-normalized revenue adjustment.

10 Results

11 **Q. What were the results of Staff's weather normalization analysis?**

12 A. Staff's weather normalization results in a volumetric adjustment of (4,174,223)
13 kWh resulting in a revenue decrease of \$(276,801). Empire's weather
14 normalization resulted in a volumetric adjustment of (2,763,500) kWh, resulting in
15 a revenue decrease of \$(195,088). Staff is proposing an adjustment of \$(81,713),
16 the difference between Staff's and Empire's results. Staff's regression results are
17 displayed in Exhibit DLP-1. Staff's weather normalization volumetric and revenue
18 adjustments are illustrated in Exhibit DLP-2.

19 Recommendation

20 **Q. Do you have a recommendation?**

21 A. Yes. Since the weather experienced in Empire's service territory during the test
22 year deviated from normal weather for that area, an adjustment is necessary to

⁸ $\left(\begin{array}{c} \text{Estimated} \\ \text{effect on revenue collection} \end{array} \right) = \left(\begin{array}{c} \text{Volumetric} \\ \text{Sales Adjustment} \end{array} \right) * \left(\begin{array}{c} \text{Applicable} \\ \text{Tariff Rate} \end{array} \right)$

1 ensure test year revenue reflects Empire's normal ongoing operations. Because
2 Staff utilized standard econometric techniques, corrected data errors, and assured
3 usage matched the month it occurred in, I recommend the Commission accept
4 Staff's weather normalization revenue adjustment of \$(81,961).

5 **CUSTOMER ANNUALIZATION**

6 **Purpose of Customer Annualization**

7 **Q. What is the purpose of annualizing customer counts?**

8 A. Customer annualization adjusts test year revenues to reflect the number of
9 customers for each customer class Empire was serving at the end of the test year.
10 Thus, the adjustment represents the revenue Empire would have received if the
11 number of customers at year-end had received service throughout the entire test
12 year.

13 **Process**

14 **Q. Please provide a brief description of the customer annualization process.**

15 A. The customer annualization process can be divided into five steps:
16 1) Collection of customer counts by weather station per customer class;
17 2) Calculation of customer coefficients;
18 3) Adjust monthly customer counts;
19 4) Calculation of volumetric adjustment; and
20 5) Calculation of revenue adjustment.
21

22 **Data Collection**

23 **Q. How did Staff obtain the customer counts per customer class and weather**
24 **station?**

25 A. Empire supplied monthly customer counts per rate class from July 2011 through
26 June 2018.

1 Customer Coefficient Calculation

2 **Q. What is a customer coefficient?**

3 A. The customer coefficient represents the change in customers each month, assuming
4 the change occurs at a constant rate throughout the test year.

5 **Q. How did you calculate the customer coefficients?**

6 A. Staff calculated customer coefficients by subtracting June 2018 customer counts
7 from June 2017 customer counts for each rate class. This value was then divided
8 by twelve to evenly spread the difference across the test-year months.⁹

9 Customer Count Adjustment Process

10 **Q. Can you describe how the customer coefficients are used to calculate the**
11 **annualized monthly customer counts?**

12 A. Yes. Beginning in July of the test year, the customer coefficient is multiplied by
13 11.5 (August by 10.5, and so on) and continues until June, when actual customer
14 count and annualized customer count are equal.

15 **Q. Why did Staff annualize customer counts using this method?**

16 A. By using this method two things are accomplished. First, it simulates the number
17 of customers Empire was serving at the end of the test year were served throughout
18 the test year. Second, by multiplying by 11.5 and so on, Staff is approximating the
19 change in the number of bills resulting from the increase/decrease of customers
20 joining at different times throughout the month instead of all at the beginning.

⁹ $Customer\ Coefficient = \frac{June\ 2018\ Customer\ Count - June\ 2017\ Customer\ Count}{12}$

1 *Volumetric Adjustment Process*

2 **Q. How did you calculate the volume adjustment?**

3 A. In order to derive annualized monthly volumes, Staff multiplied the difference
4 between annualized customer count and actual customer count times the monthly
5 weather normalized volumes per customer across each rate class and weather
6 station.

7 *Revenue Adjustment Process*

8 **Q. How did you calculate the revenue adjustment?**

9 A. In order to arrive at monthly adjusted revenues, Staff added the product of the
10 annualized monthly volumes, determined in the previous calculation, and the
11 corresponding kWh charge and the product of the annualized customer count and
12 the corresponding basic service charge. The final test year adjustment is the sum
13 of adjusted revenues across all months in the test year associated with the
14 customer annualization according to customer class and weather station.

15 **Results**

16 **Q. What customer annualization adjustment are you recommending?**

17 A. Staff's calculation results in a revenue increase of \$40,417. Empire calculated a
18 customer annualization adjustment, but did not include it in their final schedules.
19 Therefore, Staff is proposing an adjustment of \$40,417, the difference between
20 Staff and Empire's filed adjustments. Staff's customer annualization results are
21 displayed in Exhibit DLP – 3.

22 **Recommendation**

23 **Q. Do you have a recommendation?**

24 A. Yes. Weather normalized volumes are an input into the customer annualization
25 adjustment, and for the reasons mentioned above, Staff believes its weather

IV. CONCLUSION

8 A. The Commission should accept Staff’s proposed weather normalization revenue
9 adjustment of \$(81,713) and customer annualization adjustment of \$40,417 because
10 Staff utilized standard techniques, corrected data errors, and assured volumetric
11 usage matched the month it occurred in to calculate the weather normalization and
12 customer annualization adjustments.

14 A. Yes. Thank you.

EXHIBIT DLP – 1
19-EPDE-223-RTS
REGRESSION RESULTS

Residential_General				
Dependent Variable: RG_AVG_USAGE				
Method: ARMA Maximum Likelihood (BFGS)				
Date: 04/23/19 Time: 11:26				
Sample: 2011M08 2018M06				
Included observations: 83				
Convergence achieved after 6 iterations				
Coefficient covariance computed using outer product of gradients				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	351.4274	32.87894	10.68853	0
SPRG_MO_CDD	0.885755	0.084082	10.53443	0
SPRG_MO_CDD(-1)	1.542861	0.118871	12.97931	0
SPRG_MO_HDD	0.342694	0.049325	6.947713	0
SPRG_MO_HDD(-1)	0.451276	0.060806	7.421541	0
AR(12)	0.482009	0.107662	4.47707	0
SIGMASQ	3338.6	613.2803	5.443841	0
R-squared	0.950802	Mean dependent var		944.1133
Adjusted R-squared	0.946918	S.D. dependent var		262.0846
S.E. of regression	60.38297	Akaike info criterion		11.15808
Sum squared resid	277103.8	Schwarz criterion		11.36208
Log likelihood	-456.0605	Hannan-Quinn criter.		11.24004
F-statistic	244.7975	Durbin-Watson stat		2.21141
Prob(F-statistic)	0			
Inverted AR Roots	0.94	.81+.47i	.81-.47i	.47+.81i
	.47-.81i	.00-.94i	-.00+.94i	-.47-.81i
	-.47+.81i	-.81+.47i	-.81-.47i	-0.94

Residential_Water Heating				
Dependent Variable: RG_WH_AVG_USAGE_2				
Method: ARMA Maximum Likelihood (BFGS)				
Date: 04/26/19 Time: 13:01				
Sample: 2011M08 2018M06				
Included observations: 83				
Convergence achieved after 5 iterations				
Coefficient covariance computed using outer product of gradients				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	487.1142	36.8186	13.23011	0
SPRG_MO_CDD	0.888065	0.09977	8.9011	0
SPRG_MO_CDD(-1)	1.673902	0.134052	12.48699	0
SPRG_MO_HDD	0.399778	0.052559	7.606306	0
SPRG_MO_HDD(-1)	0.633881	0.069547	9.114386	0
AR(12)	0.407342	0.121286	3.358526	0.0012
SIGMASQ	4192.128	731.0562	5.734345	0
R-squared	0.944441	Mean dependent var		1182.477
Adjusted R-squared	0.940055	S.D. dependent var		276.3575
S.E. of regression	67.66273	Akaike info criterion		11.37375
Sum squared resid	347946.7	Schwarz criterion		11.57775
Log likelihood	-465.0105	Hannan-Quinn criter.		11.4557
F-statistic	215.3182	Durbin-Watson stat		2.214367
Prob(F-statistic)	0			
Inverted AR Roots	0.93	.80+.46i	.80-.46i	.46+.80i
	.46-.80i	.00+.93i	-.00-.93i	-.46+.80i
	-.46-.80i	-.80-.46i	-.80+.46i	-0.93

Residential_Total Electric				
Dependent Variable: RH_AVG_USAGE				
Method: ARMA Maximum Likelihood (OPG - BHHH)				
Date: 04/23/19 Time: 11:13				
Sample: 2011M08 2018M06				
Included observations: 83				
Convergence achieved after 13 iterations				
Coefficient covariance computed using outer product of gradients				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	348.7872	60.45709	5.76917	0
SPRG_MO_CDD	1.096615	0.264368	4.148062	0.0001
SPRG_MO_CDD(-1)	1.894473	0.239601	7.906781	0
SPRG_MO_HDD	0.886422	0.093873	9.442745	0
SPRG_MO_HDD(-1)	1.369244	0.097735	14.0098	0
AR(12)	0.327011	0.106053	3.083459	0.0029
SIGMASQ	13414.76	1959.056	6.847564	0
R-squared	0.947529	Mean dependent var		1534.212
Adjusted R-squared	0.943386	S.D. dependent var		508.7002
S.E. of regression	121.0385	Akaike info criterion		12.52701
Sum squared resid	1113425	Schwarz criterion		12.73101
Log likelihood	-512.8711	Hannan-Quinn criter.		12.60897
F-statistic	228.7343	Durbin-Watson stat		2.263272
Prob(F-statistic)	0			
Inverted AR Roots	0.91	.79+.46i	.79-.46i	.46+.79i
	.46-.79i	.00+.91i	.00-.91i	-.46-.79i
	-.46+.79i	-.79+.46i	-.79-.46i	-0.91

Commercial				
Dependent Variable: CB_AVG_USAGE				
Method: ARMA Maximum Likelihood (BFGS)				
Date: 04/26/19 Time: 15:16				
Sample: 2011M08 2018M06				
Included observations: 83				
Convergence achieved after 5 iterations				
Coefficient covariance computed using outer product of gradients				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	798.7503	31.84211	25.08472	0
SPRG_MO_CDD	0.524136	0.087133	6.015361	0
SPRG_MO_CDD(-1)	1.651005	0.079078	20.87818	0
SPRG_MO_HDD	0.183548	0.053407	3.436754	0.001
SPRG_MO_HDD(-1)	0.389699	0.051265	7.60165	0
AR(3)	0.263449	0.119645	2.201924	0.0307
SAR(12)	0.361119	0.116478	3.100305	0.0027
SIGMASQ	3714.763	692.0682	5.367626	0
R-squared	0.937875	Mean dependent var		1288.606
Adjusted R-squared	0.932077	S.D. dependent var		246.0174
S.E. of regression	64.11712	Akaike info criterion		11.27365
Sum squared resid	308325.4	Schwarz criterion		11.50679
Log likelihood	-459.8563	Hannan-Quinn criter.		11.36731
F-statistic	161.7501	Durbin-Watson stat		1.9935
Prob(F-statistic)	0			
Inverted AR Roots	0.92	.80+.46i	.80-.46i	0.64
	.46+.80i	.46-.80i	.00+.92i	.00-.92i
	-.32+.56i	-.32-.56i	-.46+.80i	-.46-.80i
	-.80-.46i	-.80+.46i	-0.92	

General Power				
Dependent Variable: GP_AVG_USAGE				
Method: ARMA Conditional Least Squares (BFGS / Marquardt steps)				
Date: 04/17/19 Time: 11:33				
Sample (adjusted): 2012M09 2018M06				
Included observations: 70 after adjustments				
Convergence achieved after 28 iterations				
HAC standard errors & covariance (Bartlett kernel, Newey-West fixed bandwidth = 4.0000)				
MA Backcast: 2011M08 2012M08				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	25386.84	631.9817	40.17022	0
SPRG_MO_CDD(-1)	21.47222	1.766962	12.15206	0
SPRG_MO_HDD	1.821109	0.840756	2.166038	0.0341
AR(1)	0.786778	0.09676	8.131256	0
SAR(12)	-0.282704	0.106474	-2.65515	0.01
MA(1)	-0.70473	0.123608	-5.701338	0
SMA(12)	0.875155	0.029989	29.18286	0
R-squared	0.869328	Mean dependent var		28602.46
Adjusted R-squared	0.856883	S.D. dependent var		3214.23
S.E. of regression	1215.97	Akaike info criterion		17.13911
Sum squared resid	93150777	Schwarz criterion		17.36396
Log likelihood	-592.8689	Hannan-Quinn criter.		17.22842
F-statistic	69.85367	Durbin-Watson stat		2.324849
Prob(F-statistic)	0			
Inverted AR Roots	.87-.23i	.87+.23i	0.79	.64+.64i
	.64-.64i	.23-.87i	.23+.87i	-.23+.87i
	-.23-.87i	-.64-.64i	-.64-.64i	-.87-.23i
	-.87+.23i			
Inverted MA Roots	.96+.26i	.96-.26i	0.7	.70+.70i
	.70-.70i	.26+.96i	.26-.96i	-.26-.96i
	-.26+.96i	-.70-.70i	-.70-.70i	-.96-.26i
	-.96+.26i			

Space Heating				
Dependent Variable: SH_AVG_USAGE				
Method: ARMA Maximum Likelihood (BFGS)				
Date: 04/17/19 Time: 13:39				
Sample: 2011M08 2018M06				
Included observations: 83				
Convergence achieved after 5 iterations				
Coefficient covariance computed using outer product of gradients				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	866.57	66.4465	13.04162	0
SPRG_MO_CDD	1.274429	0.305881	4.166421	0.0001
SPRG_MO_CDD(-1)	2.229023	0.249429	8.936519	0
SPRG_MO_HDD	0.645039	0.115862	5.567279	0
SPRG_MO_HDD(-1)	1.613741	0.112171	14.38647	0
AR(1)	-0.314807	0.117721	-2.674171	0.0092
SAR(12)	0.456289	0.108187	4.21761	0.0001
SIGMASQ	21324.86	3125.679	6.822473	0
R-squared	0.920533	Mean dependent var		2116.625
Adjusted R-squared	0.913116	S.D. dependent var		521.1717
S.E. of regression	153.6213	Akaike info criterion		13.03329
Sum squared resid	1769963	Schwarz criterion		13.26643
Log likelihood	-532.8814	Hannan-Quinn criter.		13.12695
F-statistic	124.112	Durbin-Watson stat		1.990317
Prob(F-statistic)	0			
Inverted AR Roots	0.94	.81+.47i	.81-.47i	.47+.81i
	.47-.81i	.00-.94i	.00+.94i	-0.31
	-.47-.81i	-.47+.81i	-.81-.47i	-.81+.47i
	-0.94			

Total Electric Building				
Dependent Variable: TEB_AVG_USAGE				
Method: ARMA Conditional Least Squares (Gauss-Newton / Marquardt steps)				
Date: 04/17/19 Time: 11:35				
Sample (adjusted): 2012M09 2018M06				
Included observations: 70 after adjustments				
Convergence achieved after 6 iterations				
HAC standard errors & covariance (Bartlett kernel, Newey-West fixed bandwidth = 4.0000)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	11171.22	873.2591	12.79256	0
SPRG_MO_CDD(-1)	22.00696	2.641466	8.331342	0
SPRG_MO_HDD	3.386453	0.931931	3.633804	0.0006
SPRG_MO_HDD(-1)	11.46111	1.222091	9.378277	0
TEB_DUMMY	-9261.831	521.2008	-17.77018	0
AR(1)	0.328476	0.090485	3.630182	0.0006
SAR(12)	0.242041	0.068236	3.547129	0.0007
R-squared	0.864414	Mean dependent var	18872.61	
Adjusted R-squared	0.851501	S.D. dependent var	4059.347	
S.E. of regression	1564.292	Akaike info criterion	17.64289	
Sum squared resid	1.54E+08	Schwarz criterion	17.86774	
Log likelihood	-610.5013	Hannan-Quinn criter.	17.73221	
F-statistic	66.94166	Durbin-Watson stat	1.997298	
Prob(F-statistic)	0			
Inverted AR Roots	0.89	.77+.44i	.77-.44i	.44-.77i
	.44+.77i	0.33	.00+.89i	.00-.89i
	-.44+.77i	-.44-.77i	-.77-.44i	-.77+.44i
	-0.89			

EXHIBIT DLP - 2

19-EPDE-223-RTS

WEATHER NORMALIZATION ADJUSTMENTS

Weather Normalization Summary			
Customer Classification		Staff's Volumetric Adjustment	Staff's Revenue Adjustment
RG	RG	(2,346,077)	\$ (152,747)
RG-WH	RG-WH	(306,044)	\$ (19,003)
RH	RH	(687,076)	\$ (39,322)
CB	CB	(410,818)	\$ (35,191)
GP	GP	(312,723)	\$ (22,652)
SH	SH	(52,293)	\$ (3,780)
TEB	TEB	(59,194)	\$ (4,107)
Total		(4,174,223)	\$ (276,801)

Exhibit DLP – 2
19-EPDE-223-RTS
Weather Normalization Adjustment

Residential General																			
Weather Station	Customer Classification	TY: 201707 - 201806 Month	Customer Count	Usage	Average Usage	Basic Service	kWh Usage Rate	Current Month HDD Actual	Previous Month HDD Normal	Current Month CDD Actual	Previous Month CDD Normal	Change in Usage due to HDD	Change in Usage due to CDD	Total Adjustment	Revenue Adjustment				
SPRG	RG	June	5,536					0.342694	0.451276		0.885755	1.542861							
		July	5,538	6,649,942	1200.78	\$ 14.00	\$ 0.0643	0.0	0.6	0.0	7.3	457.0	412.1	263.5	277.5	3.50	(18.13)	(81,032)	\$ (5,209)
		August	5,537	7,119,677	1285.84	\$ 14.00	\$ 0.0641	0.0	1.5	0.0	0.6	268.0	391.4	457.0	412.1	0.79	39.97	225,695	\$ 14,468
		September	5,539	5,499,767	992.92	\$ 14.00	\$ 0.0648	8.5	47.1	0.0	1.5	211.0	171.3	268.0	391.4	13.91	155.14	936,411	\$ 60,717
		October	5,534	4,443,856	803.01	\$ 14.00	\$ 0.0655	237.5	250.1	8.5	47.1	84.0	32.4	211.0	171.3	21.73	(107.04)	(472,093)	\$ (30,927)
		November	5,540	3,878,852	700.15	\$ 14.00	\$ 0.0659	453.0	552.2	237.5	250.1	3.5	1.6	84.0	32.4	39.66	(81.32)	(230,754)	\$ (15,200)
		December	5,553	4,618,257	831.67	\$ 14.00	\$ 0.0652	878.0	896.0	453.0	552.2	0.0	0.1	3.5	1.6	50.91	(2.84)	266,970	\$ 17,414
		January	5,547	7,038,685	1268.92	\$ 14.00	\$ 0.0640	1031.5	963.8	878.0	896.0	0.0	0.0	0.0	0.1	(15.07)	0.08	(83,175)	\$ (5,321)
		February	5,562	5,849,480	1051.69	\$ 14.00	\$ 0.0645	720.5	772.3	1031.5	963.8	0.5	0.3	0.0	0.0	(12.80)	(0.15)	(72,004)	\$ (4,641)
		March	5,559	4,541,451	816.95	\$ 14.00	\$ 0.0652	546.0	566.3	720.5	772.3	0.0	4.6	0.5	0.3	30.31	3.79	189,539	\$ 12,362
		April	5,537	4,054,152	732.19	\$ 14.00	\$ 0.0656	433.5	279.0	546.0	566.3	12.0	24.7	0.0	4.6	(43.81)	18.32	(141,134)	\$ (9,257)
		May	5,530	3,916,243	708.18	\$ 14.00	\$ 0.0659	5.5	99.7	433.5	279.0	263.5	99.1	12.0	24.7	(37.45)	(126.02)	(903,959)	\$ (59,544)
		June	5,554	6,225,912	1120.98	\$ 14.00	\$ 0.0644	0.0	7.3	5.5	99.7	444.5	277.5	263.5	99.1	45.02	(401.62)	(1,980,542)	\$ (127,609)
		Total			5,544	63,836,274												(2,346,077)	\$ (152,747)

Residential Water Heating																			
Weather Station	Customer Classification	TY: 201707 - 201806 Month	Customer Count	Usage	Average Usage	Basic Service	Commodity Rate	Current Month HDD Actual	Current Month HDD Normal	Previous Month HDD Actual	Previous Month HDD Normal	Current Month CDD Actual	Current Month CDD Normal	Previous Month CDD Actual	Previous Month CDD Normal	Change in Usage due to HDD	Change in Usage due to CDD	Total Adjustment	Revenue Adjustment
SPRG	RG-WH	June	765					0.399778		0.633881		0.888065		1.673902					
		July	770	1,035,706	1345.07	\$ 14.00	\$ 0.0619	0.0	0.6	0.0	7.3	457.0	412.1	263.5	277.5	4.87	(16.40)	(8,876)	\$ (549)
		August	768	1,137,109	1480.61	\$ 14.00	\$ 0.0618	0.0	1.5	0.0	0.6	268.0	391.4	457.0	412.1	0.98	34.37	27,152	\$ 1,679
		September	766	902,628	1178.37	\$ 14.00	\$ 0.0620	8.5	47.1	0.0	1.5	211.0	171.3	268.0	391.4	16.40	171.22	143,713	\$ 8,911
		October	762	745,574	978.44	\$ 14.00	\$ 0.0622	237.5	250.1	8.5	47.1	84.0	32.4	211.0	171.3	29.49	(112.36)	(63,146)	\$ (3,925)
		November	766	699,314	912.94	\$ 14.00	\$ 0.0622	453.0	552.2	237.5	250.1	3.5	1.6	84.0	32.4	47.62	(88.09)	(30,994)	\$ (1,928)
		December	766	840,251	1096.93	\$ 14.00	\$ 0.0620	878.0	896.0	453.0	552.2	0.0	0.1	3.5	1.6	70.05	(3.08)	51,295	\$ 3,182
		January	762	1,260,147	1653.74	\$ 14.00	\$ 0.0618	1031.5	963.8	878.0	896.0	0.0	0.0	0.0	0.1	(15.65)	0.08	(11,860)	\$ (733)
		February	761	1,084,828	1425.53	\$ 14.00	\$ 0.0618	720.5	772.3	1031.5	963.8	0.5	0.3	0.0	0.0	(22.20)	(0.15)	(17,008)	\$ (1,052)
		March	766	851,489	1111.60	\$ 14.00	\$ 0.0620	546.0	566.3	720.5	772.3	0.0	4.6	0.5	0.3	40.92	3.78	34,238	\$ 2,123
		April	759	738,443	972.92	\$ 14.00	\$ 0.0621	433.5	279.0	546.0	566.3	12.0	24.7	0.0	4.6	(48.94)	18.95	(22,758)	\$ (1,414)
		May	745	695,303	933.29	\$ 14.00	\$ 0.0622	5.5	99.7	433.5	279.0	263.5	99.1	12.0	24.7	(60.29)	(124.73)	(137,835)	\$ (8,574)
		June	748	940,759	1257.70	\$ 14.00	\$ 0.0620	0.0	7.3	5.5	99.7	444.5	277.5	263.5	99.1	62.64	(423.55)	(269,964)	\$ (16,725)
		Total		762	10,931,551														(306,044)

Exhibit DLP – 2
19-EPDE-223-RTS
Weather Normalization Adjustment

Residential Total Electric																						
Weather	Customer	TY: 201707 - 201806	Customer	Usage	Average Usage	Basic	Commodity	Current Month HDD		Previous Month HDD		Current Month CDD		Previous Month CDD		Change in Usage	Change in Usage	Total	Revenue			
Station	Classification	Month	Count			Service	Rate	Actual	Normal	Actual	Normal	Actual	Normal	Actual	Normal	due to HDD	due to CDD	Adjustment	Adjustment			
SPRG																						
		June	1,857					0.886422		1.369244		1.096615		1.894473								
	RH	July	1,855	2,561,730	1380.99	\$	14.00	\$	0.0572	0.0	0.6	0.0	7.3	457.0	412.1	263.5	277.5	10.54	(22.67)	(22,511)	\$	(1,288)
		August	1,862	2,733,690	1468.15	\$	14.00	\$	0.0572	0.0	1.5	0.0	0.6	268.0	391.4	457.0	412.1	2.16	50.19	97,476	\$	5,579
		September	1,860	2,252,828	1211.20	\$	14.00	\$	0.0572	8.5	47.1	0.0	1.5	211.0	171.3	268.0	391.4	36.30	190.14	421,184	\$	24,104
		October	1,868	1,899,621	1016.93	\$	14.00	\$	0.0572	237.5	250.1	8.5	47.1	84.0	32.4	211.0	171.3	64.00	(131.90)	(126,831)	\$	(7,259)
		November	1,865	2,241,550	1201.90	\$	14.00	\$	0.0572	453.0	552.2	237.5	250.1	3.5	1.6	84.0	32.4	105.14	(99.86)	9,840	\$	563
		December	1,871	2,892,370	1545.90	\$	14.00	\$	0.0572	878.0	896.0	453.0	552.2	0.0	0.1	3.5	1.6	151.72	(3.48)	277,347	\$	15,873
		January	1,868	5,297,095	2835.70	\$	14.00	\$	0.0572	1031.5	963.8	878.0	896.0	0.0	0.0	0.0	0.1	(35.35)	0.09	(65,856)	\$	(3,769)
		February	1,873	4,394,934	2346.47	\$	14.00	\$	0.0572	720.5	772.3	1031.5	963.8	0.5	0.3	0.0	0.0	(46.77)	(0.18)	(87,948)	\$	(5,033)
		March	1,873	3,191,208	1703.79	\$	14.00	\$	0.0572	546.0	566.3	720.5	772.3	0.0	4.6	0.5	0.3	88.85	4.69	175,212	\$	10,027
		April	1,866	2,657,124	1423.97	\$	14.00	\$	0.0572	433.5	279.0	546.0	566.3	12.0	24.7	0.0	4.6	(109.24)	22.61	(161,642)	\$	(9,251)
		May	1,869	2,102,807	1125.10	\$	14.00	\$	0.0572	5.5	99.7	433.5	279.0	263.5	99.1	12.0	24.7	(128.07)	(156.22)	(531,330)	\$	(30,408)
		June	1,871	2,394,743	1279.93	\$	14.00	\$	0.0572	0.0	7.3	5.5	99.7	444.5	277.5	263.5	99.1	135.47	(494.64)	(672,018)	\$	(38,460)
		Total	1,867	34,619,700																(687,076)	\$	(39,322)

Commercial																			
Weather	Customer	TY: 201707 - 201806	Customer	Usage	Average Usage	Basic	Commodity	Current Month HDD		Previous Month HDD		Current Month CDD		Previous Month CDD		Change in Usage	Change in Usage	Total	Revenue
Station	Classification	Month	Count			Service	Rate	Actual	Normal	Actual	Normal	Actual	Normal	Actual	Normal	due to HDD	due to CDD	Adjustment	Adjustment
SPRG																			
		June	1,187					0.183548		0.389699		0.524136		1.651005					
	CB	July	1,191	1,710,904	1436.53	\$ 19.00	\$ 0.0853	0.0	0.6	0.0	7.3	457.0	412.1	263.5	277.5	2.96	(0.37)	3,079	\$ 263
		August	1,188	1,914,908	1611.88	\$ 19.00	\$ 0.0851	0.0	1.5	0.0	0.6	268.0	391.4	457.0	412.1	0.51	(9.50)	(10,678)	\$ (909)
		September	1,187	1,738,104	1464.28	\$ 19.00	\$ 0.0852	8.5	47.1	0.0	1.5	211.0	171.3	268.0	391.4	7.68	182.85	226,162	\$ 19,279
		October	1,184	1,485,993	1255.06	\$ 19.00	\$ 0.0856	237.5	250.1	8.5	47.1	84.0	32.4	211.0	171.3	17.35	(92.66)	(89,172)	\$ (7,632)
		November	1,184	1,244,588	1051.17	\$ 19.00	\$ 0.0861	453.0	552.2	237.5	250.1	3.5	1.6	84.0	32.4	23.11	(86.23)	(74,730)	\$ (6,432)
		December	1,182	1,418,549	1200.13	\$ 19.00	\$ 0.0857	878.0	896.0	453.0	552.2	0.0	0.1	3.5	1.6	41.94	(3.06)	45,964	\$ 3,940
		January	1,184	1,555,313	1313.61	\$ 19.00	\$ 0.0857	1031.5	963.8	878.0	896.0	0.0	0.0	0.0	0.1	(5.41)	0.08	(6,306)	\$ (541)
		February	1,181	1,670,255	1414.27	\$ 19.00	\$ 0.0853	720.5	772.3	1031.5	963.8	0.5	0.3	0.0	0.0	(16.87)	(0.09)	(20,028)	\$ (1,709)
		March	1,181	1,246,764	1055.69	\$ 19.00	\$ 0.0863	546.0	566.3	720.5	772.3	0.0	4.6	0.5	0.3	23.90	2.12	30,724	\$ 2,653
		April	1,181	1,224,556	1036.88	\$ 19.00	\$ 0.0862	433.5	279.0	546.0	566.3	12.0	24.7	0.0	4.6	(20.47)	14.21	(7,389)	\$ (637)
		May	1,188	1,337,589	1125.92	\$ 19.00	\$ 0.0858	5.5	99.7	433.5	279.0	263.5	99.1	12.0	24.7	(42.92)	(65.17)	(128,418)	\$ (11,024)
		June	1,184	1,635,311	1381.17	\$ 19.00	\$ 0.0854	0.0	7.3	5.5	99.7	444.5	277.5	263.5	99.1	38.05	(359.02)	(380,027)	\$ (32,442)
		Total	1,185	18,182,834														(410,818)	\$ (35,191)

Exhibit DLP – 2
19-EPDE-223-RTS
Weather Normalization Adjustment

General Power																		
Weather	Customer	TY: 201707 - 201806	Customer	Usage	Average Usage	Basic	Commodity	Current Month HDD	Previous Month HDD	Current Month CDD	Previous Month CDD	Change in Usage	Change in Usage	Total	Revenue			
Station	Classification	Month	Count			Service	Rate	Actual	Normal	Actual	Normal	Actual	Normal	due to HDD	due to CDD	Adjustment	Adjustment	
SPRG																		
	GP	June	104					1.821109		0		0		21.47222				
		July	105	3,300,387	31432.26	\$ -	\$ 0.0718	0.0	0.6	0.0	7.3	457.0	412.1	263.5	277.5	1.06	301.33	31,751 \$ 2,280
		August	105	3,578,020	34076.38	\$ -	\$ 0.0700	0.0	1.5	0.0	0.6	268.0	391.4	457.0	412.1	2.79	(964.46)	(100,975) \$ (7,068)
		September	106	3,630,302	34248.13	\$ -	\$ 0.0699	8.5	47.1	0.0	1.5	211.0	171.3	268.0	391.4	70.26	2,648.96	288,237 \$ 20,148
		October	106	3,188,348	30078.75	\$ -	\$ 0.0746	237.5	250.1	8.5	47.1	84.0	32.4	211.0	171.3	22.95	(853.16)	(88,003) \$ (6,564)
		November	106	3,042,124	28699.28	\$ -	\$ 0.0744	453.0	552.2	237.5	250.1	3.5	1.6	84.0	32.4	180.56	(1,108.68)	(98,381) \$ (7,316)
		December	106	2,855,219	26936.03	\$ -	\$ 0.0754	878.0	896.0	453.0	552.2	0.0	0.1	3.5	1.6	32.78	(40.08)	(774) \$ (58)
		January	106	3,164,409	29852.92	\$ -	\$ 0.0728	1031.5	963.8	878.0	896.0	0.0	0.0	0.0	0.1	(123.26)	1.07	(12,952) \$ (943)
		February	106	2,868,968	27065.74	\$ -	\$ 0.0732	720.5	772.3	1031.5	963.8	0.5	0.3	0.0	0.0	94.30	-	9,996 \$ 731
		March	106	2,788,196	26303.74	\$ -	\$ 0.0744	546.0	566.3	720.5	772.3	0.0	4.6	0.5	0.3	36.88	(3.58)	3,530 \$ 263
		April	106	2,608,320	24606.79	\$ -	\$ 0.0763	433.5	279.0	546.0	566.3	12.0	24.7	0.0	4.6	(281.39)	98.06	(19,434) \$ (1,483)
		May	106	2,884,387	27211.20	\$ -	\$ 0.0756	5.5	99.7	433.5	279.0	263.5	99.1	12.0	24.7	171.55	273.41	47,166 \$ 3,565
		June	106	3,427,680	32336.60	\$ -	\$ 0.0703	0.0	7.3	5.5	99.7	444.5	277.5	263.5	99.1	13.32	(3,531.11)	(372,885) \$ (26,206)
		Total		106	37,336,360													(312,723) \$ (22,652)

Space Heating																			
Weather Station	Customer Classification	TY: 201707 - 201806 Month	Customer Count	Usage	Average Usage	Basic Service	Commodity Rate	Current Month HDD Actual	Current Month HDD Normal	Previous Month HDD Actual	Previous Month HDD Normal	Current Month CDD Actual	Current Month CDD Normal	Previous Month CDD Actual	Previous Month CDD Normal	Change in Usage due to HDD	Change in Usage due to CDD	Total Adjustment	Revenue Adjustment
SPRG																			
	SH	June	110					0.645039		1.613741		1.274429		2.229023					
		July	109	243,165	2230.87	\$ 19.00	\$ 0.0725	0.0	0.6	0.0	7.3	457.0	412.1	263.5	277.5	12.18	(25.96)	(1,502)	\$ (109)
		August	110	230,398	2094.53	\$ 19.00	\$ 0.0727	0.0	1.5	0.0	0.6	268.0	391.4	457.0	412.1	1.93	57.10	6,494	\$ 472
		September	110	210,731	1915.74	\$ 19.00	\$ 0.0729	8.5	47.1	0.0	1.5	211.0	171.3	268.0	391.4	27.36	224.35	27,688	\$ 2,019
		October	110	179,771	1634.28	\$ 19.00	\$ 0.0733	237.5	250.1	8.5	47.1	84.0	32.4	211.0	171.3	70.39	(154.37)	(9,238)	\$ (677)
		November	110	174,712	1588.29	\$ 19.00	\$ 0.0735	453.0	552.2	237.5	250.1	3.5	1.6	84.0	32.4	84.29	(117.47)	(3,650)	\$ (268)
		December	109	211,027	1936.03	\$ 19.00	\$ 0.0730	878.0	896.0	453.0	552.2	0.0	0.1	3.5	1.6	171.61	(4.10)	18,259	\$ 1,334
		January	109	338,183	3102.60	\$ 19.00	\$ 0.0719	1031.5	963.8	878.0	896.0	0.0	0.0	0.0	0.1	(14.61)	0.11	(1,580)	\$ (114)
		February	110	316,848	2880.44	\$ 19.00	\$ 0.0721	720.5	772.3	1031.5	963.8	0.5	0.3	0.0	0.0	(75.82)	(0.21)	(8,364)	\$ (603)
		March	110	230,202	2092.75	\$ 19.00	\$ 0.0729	546.0	566.3	720.5	772.3	0.0	4.6	0.5	0.3	96.63	5.45	11,228	\$ 819
		April	110	207,257	1884.15	\$ 19.00	\$ 0.0735	433.5	279.0	546.0	566.3	12.0	24.7	0.0	4.6	(66.99)	26.41	(4,464)	\$ (328)
		May	110	270,808	2461.89	\$ 19.00	\$ 0.0722	5.5	99.7	433.5	279.0	263.5	99.1	12.0	24.7	(188.59)	(181.20)	(40,676)	\$ (2,935)
		June	110	210,616	1914.69	\$ 19.00	\$ 0.0729	0.0	7.3	5.5	99.7	444.5	277.5	263.5	99.1	156.73	(579.35)	(46,488)	\$ (3,389)
		Total		110	2,823,718														(52,293)

Exhibit DLP – 2
19-EPDE-223-RTS
Weather Normalization Adjustment

Total Electric Building																			
Weather Station	Customer Classification	TY: 201707 - 201806 Month	Customer Count	Usage	Avg Usage	Basic Service	Commodity Rate	Current Month HDD Actual	Current Month HDD Normal	Previous Month HDD Actual	Previous Month HDD Normal	Current Month CDD Actual	Current Month CDD Normal	Previous Month CDD Actual	Previous Month CDD Normal	Change in Usage due to HDD	Change in Usage due to CDD	Total Adjustment	Revenue Adjustment
SPRG																			
		June	39					3.386453	11.46111			0	22.00696						
	TEB	July	39	724,582	18579.03	\$ -	\$ 0.0710	0.0	0.6	0.0	7.3	457.0	412.1	263.5	277.5	85.83	308.83	15,392	\$ 1,094
		August	39	828,185	21235.51	\$ -	\$ 0.0698	0.0	1.5	0.0	0.6	268.0	391.4	457.0	412.1	11.88	(988.48)	(38,087)	\$ (2,659)
		September	39	774,232	19852.10	\$ -	\$ 0.0704	8.5	47.1	0.0	1.5	211.0	171.3	268.0	391.4	148.23	2,714.93	111,663	\$ 7,864
		October	39	728,596	18681.95	\$ -	\$ 0.0706	237.5	250.1	8.5	47.1	84.0	32.4	211.0	171.3	484.88	(874.41)	(15,192)	\$ (1,072)
		November	39	628,377	16112.23	\$ -	\$ 0.0722	453.0	552.2	237.5	250.1	3.5	1.6	84.0	32.4	480.18	(1,136.29)	(25,589)	\$ (1,847)
		December	40	711,019	17775.48	\$ -	\$ 0.0709	878.0	896.0	453.0	552.2	0.0	0.1	3.5	1.6	1,197.33	(41.08)	46,250	\$ 3,280
		January	40	1,041,092	26027.30	\$ -	\$ 0.0678	1031.5	963.8	878.0	896.0	0.0	0.0	0.0	0.1	(22.91)	1.10	(872)	\$ (59)
		February	40	1,054,332	26358.30	\$ -	\$ 0.0675	720.5	772.3	1031.5	963.8	0.5	0.3	0.0	0.0	(600.36)	-	(24,015)	\$ (1,620)
		March	40	816,305	20407.63	\$ -	\$ 0.0695	546.0	566.3	720.5	772.3	0.0	4.6	0.5	0.3	662.07	(3.67)	26,336	\$ 1,831
		April	40	702,877	17571.93	\$ -	\$ 0.0706	433.5	279.0	546.0	566.3	12.0	24.7	0.0	4.6	(291.18)	100.50	(7,627)	\$ (538)
		May	40	685,777	17144.43	\$ -	\$ 0.0705	5.5	99.7	433.5	279.0	263.5	99.1	12.0	24.7	(1,451.93)	280.22	(46,868)	\$ (3,304)
		June	40	740,732	18518.30	\$ -	\$ 0.0703	0.0	7.3	5.5	99.7	444.5	277.5	263.5	99.1	1,104.41	(3,619.04)	(100,585)	\$ (7,076)
		Total	40	9,436,106														(59,194)	\$ (4,107)

EXHIBIT DLP – 3
19-EPDE-223-RTS
CUSTOMER ANNUALIZATION ADJUSTMENT

Customer Annualization Summary			
Customer Classification		Staff's Volumetric Adjustment	Staff's Revenue Adjustment
RG	RG	\$ 109,323	\$ 8,584
RG-WH	RG-WH	\$ (126,439)	\$ (9,262)
RH	RH	\$ 129,037	\$ 8,561
CB	CB	\$ (24,124)	\$ (2,404)
GP	GP	\$ 365,108	\$ 26,473
SH	SH	\$ -	\$ -
TEB	TEB	\$ 120,772	\$ 8,465
		<u>\$ 573,676.43</u>	<u>\$ 40,416.56</u>

Exhibit DLP - 3
19-EPDE-223-RTS
Customer Annualization Adjustment

Residential General						
Customer Coefficient	Annualized Customer Count	Total Volumetric Adjustment	Volumetric Revenue Adjustment	Customer Charge Revenue Adjustment	Total Revenue Adjustment	
1.50	17	20,461	\$ 1,315	\$ 242	\$ 1,557	
1.50	16	20,894	\$ 1,339	\$ 221	\$ 1,560	
1.50	14	16,558	\$ 1,074	\$ 200	\$ 1,273	
1.50	13	9,151	\$ 599	\$ 179	\$ 778	
1.50	11	7,408	\$ 488	\$ 158	\$ 645	
1.50	10	8,578	\$ 559	\$ 137	\$ 696	
1.50	8	10,345	\$ 662	\$ 116	\$ 777	
1.50	7	7,012	\$ 452	\$ 95	\$ 546	
1.50	5	4,468	\$ 291	\$ 74	\$ 365	
1.50	4	2,650	\$ 174	\$ 53	\$ 226	
1.50	2	1,226	\$ 81	\$ 32	\$ 112	
1.50	1	573	\$ 37	\$ 11	\$ 47	
	<u>9</u>	<u>109,323</u>			<u>\$ 8,584</u>	

Residential Water Heating						
Customer Coefficient	Annualized Customer Count	Total Volumetric Adjustment	Volumetric Revenue Adjustment	Customer Charge Revenue Adjustment	Total Revenue Adjustment	
(1)	(16)	(21,726)	\$ (1,345)	\$ (228)	\$ (1,573)	
(1)	(15)	(22,550)	\$ (1,394)	\$ (208)	\$ (1,603)	
(1)	(13)	(18,384)	\$ (1,140)	\$ (188)	\$ (1,328)	
(1)	(12)	(10,784)	\$ (670)	\$ (169)	\$ (839)	
(1)	(11)	(9,270)	\$ (577)	\$ (149)	\$ (725)	
(1)	(9)	(10,718)	\$ (665)	\$ (129)	\$ (794)	
(1)	(8)	(12,764)	\$ (788)	\$ (109)	\$ (897)	
(1)	(6)	(8,945)	\$ (553)	\$ (89)	\$ (642)	
(1)	(5)	(5,733)	\$ (356)	\$ (69)	\$ (425)	
(1)	(4)	(3,340)	\$ (207)	\$ (50)	\$ (257)	
(1)	(2)	(1,590)	\$ (99)	\$ (30)	\$ (129)	
(1)	(1)	(635)	\$ (39)	\$ (10)	\$ (49)	
	<u>(9)</u>	<u>(126,439)</u>			<u>\$ (9,262)</u>	

Exhibit DLP - 3
19-EPDE-223-RTS
Customer Annualization Adjustment

Residential Total Electric						
Customer Coefficient	Annualized Customer Count	Total Volumetric Adjustment	Volumetric Revenue Adjustment	Customer Charge Revenue Adjustment	Total Revenue Adjustment	
1.17	13	18,365	\$ 1,051	\$ 188	\$ 1,239	
1.17	12	18,626	\$ 1,066	\$ 172	\$ 1,237	
1.17	11	15,934	\$ 912	\$ 155	\$ 1,067	
1.17	10	9,411	\$ 539	\$ 139	\$ 677	
1.17	9	10,563	\$ 605	\$ 123	\$ 727	
1.17	8	12,847	\$ 735	\$ 106	\$ 841	
1.17	6	17,970	\$ 1,028	\$ 90	\$ 1,118	
1.17	5	12,072	\$ 691	\$ 74	\$ 764	
1.17	4	7,339	\$ 420	\$ 57	\$ 477	
1.17	3	3,901	\$ 223	\$ 41	\$ 264	
1.17	2	1,471	\$ 84	\$ 25	\$ 109	
1.17	1	537	\$ 31	\$ 8	\$ 39	
	<u>7</u>	<u>129,037</u>			<u>\$ 8,561</u>	

Commercial						
Customer Coefficient	Annualized Customer Count	Total Volumetric Adjustment	Volumetric Revenue Adjustment	Customer Charge Revenue Adjustment	Total Revenue Adjustment	
(0)	(3)	(4,137)	\$ (353)	\$ (55)	\$ (408)	
(0)	(3)	(4,208)	\$ (358)	\$ (50)	\$ (408)	
(0)	(2)	(3,930)	\$ (335)	\$ (45)	\$ (380)	
(0)	(2)	(2,507)	\$ (215)	\$ (40)	\$ (255)	
(0)	(2)	(1,853)	\$ (159)	\$ (36)	\$ (195)	
(0)	(2)	(2,013)	\$ (173)	\$ (31)	\$ (203)	
(0)	(1)	(1,799)	\$ (154)	\$ (26)	\$ (180)	
(0)	(1)	(1,572)	\$ (134)	\$ (21)	\$ (156)	
(0)	(1)	(946)	\$ (82)	\$ (17)	\$ (98)	
(0)	(1)	(644)	\$ (56)	\$ (12)	\$ (67)	
(0)	(0)	(382)	\$ (33)	\$ (7)	\$ (40)	
(0)	(0)	(133)	\$ (11)	\$ (2)	\$ (14)	
	<u>(2)</u>	<u>(24,124)</u>			<u>\$ (2,404)</u>	

Exhibit DLP - 3
19-EPDE-223-RTS
Customer Annualization Adjustment

General Power						
Customer Coefficient	Annualized Customer Count	Total Volumetric Adjustment	Volumetric Revenue Adjustment	Customer Charge Revenue Adjustment	Total Revenue Adjustment	
0.17	2	60,825	\$ 4,367	\$ -	\$ 4,367	
0.17	2	57,951	\$ 4,056	\$ -	\$ 4,056	
0.17	2	58,532	\$ 4,091	\$ -	\$ 4,091	
0.17	1	41,435	\$ 3,091	\$ -	\$ 3,091	
0.17	1	34,714	\$ 2,582	\$ -	\$ 2,582	
0.17	1	29,173	\$ 2,200	\$ -	\$ 2,200	
0.17	1	27,253	\$ 1,984	\$ -	\$ 1,984	
0.17	1	20,370	\$ 1,490	\$ -	\$ 1,490	
0.17	1	15,363	\$ 1,143	\$ -	\$ 1,143	
0.17	0	10,176	\$ 777	\$ -	\$ 777	
0.17	0	6,914	\$ 523	\$ -	\$ 523	
0.17	0	2,402	\$ 169	\$ -	\$ 169	
	<u>1</u>	<u>365,108</u>			<u>\$ 26,473</u>	

Space Heating						
Customer Coefficient	Annualized Customer Count	Total Volumetric Adjustment	Volumetric Revenue Adjustment	Customer Charge Revenue Adjustment	Total Revenue Adjustment	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
0.00	0	0	\$ -	\$ -	\$ -	
	<u>0</u>	<u>0</u>			<u>\$ -</u>	

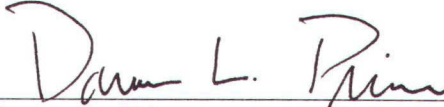
Exhibit DLP - 3
19-EPDE-223-RTS
Customer Annualization Adjustment

Total Electric Building					
Customer Coefficient	Annualized Customer Count	Total Volumetric Adjustment	Volumetric Revenue Adjustment	Customer Charge Revenue Adjustment	Total Revenue Adjustment
0.08	1	18,183	\$ 1,292	\$ -	\$ 1,292
0.08	1	17,727	\$ 1,238	\$ -	\$ 1,238
0.08	1	17,983	\$ 1,267	\$ -	\$ 1,267
0.08	1	12,957	\$ 915	\$ -	\$ 915
0.08	1	9,660	\$ 697	\$ -	\$ 697
0.08	1	10,255	\$ 727	\$ -	\$ 727
0.08	0	11,919	\$ 808	\$ -	\$ 808
0.08	0	9,659	\$ 652	\$ -	\$ 652
0.08	0	6,144	\$ 427	\$ -	\$ 427
0.08	0	3,621	\$ 256	\$ -	\$ 256
0.08	0	1,997	\$ 141	\$ -	\$ 141
0.08	0	667	\$ 47	\$ -	\$ 47
	<u>1</u>	<u>120,772</u>	8,465		<u>\$ 8,465</u>

STATE OF KANSAS)
) ss.
COUNTY OF SHAWNEE)

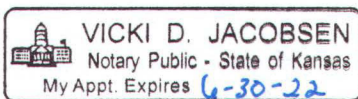
VERIFICATION

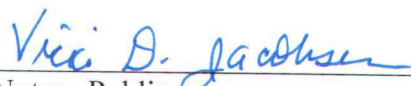
Darren L. Prince, being duly sworn upon his oath deposes and states that he is a Managing Economist for the Utilities Division of the Kansas Corporation Commission of the State of Kansas, that he has read and is familiar with the foregoing *Direct Testimony*, and attests that the statements contained therein are true and correct to the best of his knowledge, information and belief.



Darren L. Prince
Managing Economist
State Corporation Commission of the
State of Kansas

Subscribed and sworn to before me this 13th day of May, 2019.





Notary Public

My Appointment Expires: 6-30-22

CERTIFICATE OF SERVICE

19-EPDE-223-RTS

I, the undersigned, certify that a true and correct copy of the above and foregoing Direct Testimony was served via electronic service this 13th day of May, 2019, to the following:

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CERTIFICATE OF SERVICE

19-EPDE-223-RTS

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/s/ Vicki Jacobsen

Vicki Jacobsen