

**STATE OF ILLINOIS**  
**ILLINOIS COMMERCE COMMISSION**

North Shore Gas Company	:	
	:	No. 14-_____
Proposed General Increase	:	
In Rates For Gas Service	:	

Direct Direct Testimony of

**KEVIN R. KUSE**

Senior Load Forecaster  
Integrus Business Support, LLC

On Behalf of  
North Shore Gas Company

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1 **I. INTRODUCTION AND BACKGROUND**

2 **A. Identification of Witness**

3 **Q. Please state your name and business address.**

4 A. My name is Kevin R. Kuse. My business address is Integrys Energy Group, Inc.  
5 (“Integrys”), 700 North Adams Street, P.O. Box 19001, Green Bay, WI 54307-9001.

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

7 A. I am a Senior Load Forecaster in the Budgets and Forecasts Department of Integrys  
8 Business Support, LLC (“IBS”), a wholly-owned subsidiary of Integrys.

9 **Q. For whom are you providing testimony?**

10 A. I am providing testimony for North Shore Gas Company (“North Shore” or “NS”), which  
11 is a wholly-owned indirect subsidiary of Integrys.

12 **B. Purpose of Testimony and Summary of Conclusions**

13 **Q. Mr. Kuse, what is the purpose of your testimony?**

14 A. The purpose of my testimony is to present North Shore’s customer demand forecast for  
15 the 2015 test year, and to explain how that forecast was derived. I will also compare  
16 North Shore’s forecasted 2015 test year demand to its actual weather normalized demand  
17 in its last comparative year, which consists of 2013 actual weather normalized demand  
18 from January 2013 to June 2013 and forecasted demand from July 2013 to December  
19 2013 (hereinafter referred to as “comparative year 2013”). Based on its analyses, North  
20 Shore forecasts a 2015 total demand, including company use, of 34.0 billion cubic feet  
21 (“Bcf”) of natural gas as compared to 34.9 Bcf in comparative year 2013, a decrease of  
22 approximately 0.9 Bcf or about 2.6%.

23 **C. Background and Experience**

24 **Q. Please briefly outline your educational background.**

25 A. I hold a Bachelor of Arts Degree in Economics and a Master of Science Degree in  
26 Administrative Science, both from the University of Wisconsin – Green Bay.

27 **Q. Please summarize your business experience.**

28 A. In February 1993, I was hired by St. Norbert College in De Pere, Wisconsin as the  
29 Director of Research and Records in the Office of Institutional Advancement. In  
30 September 1996, I was hired as a Business Evaluation Analyst by the Development  
31 Division of the Oneida Indian Tribe of Wisconsin. In September 1999, I was hired by  
32 Wisconsin Public Service Corporation, a wholly-owned subsidiary of Integrys, as a  
33 Customer Research Analyst in the Market Research Department. From September 1999  
34 to July 2007, I developed customer insights by gathering and interpreting data from  
35 primary survey research and secondary data sources. During that period, I also  
36 performed two short term assignments as the Leader of the Market Research department.  
37 In July 2007, I became a Senior Load Forecaster in the Budgets and Forecasts  
38 Department of IBS.

39 **Q. What are your current duties and responsibilities?**

40 A. As a Senior Load Forecaster, my duties include the performance of various aspects of  
41 short-term and long-term electric and gas forecasts.

42 **Q. Have you previously testified before any regulatory agency?**

43 A. Yes, I have. I testified before the Illinois Commerce Commission (“Commission” or  
44 “ICC”) in Docket Nos. 12-0511/12-0512 (cons.), and Docket Nos. 11-0280/11-0281

45 (cons.), which were North Shore’s last two general rate cases. I have also testified before  
46 the Michigan Public Service Commission in Case No. U-16166, which was the 2011 test  
47 year rate case of Upper Peninsula Power Company, another Integrys subsidiary.

48 **II. GAS SALES FORECAST METHODOLOGY**

49 **A. Forecast of Customer Demand**

50 **Q. In general, how did North Shore forecast customer demand for the 2015 test year?**

51 A. North Shore did so by performing regression analyses for each Service Classification  
52 (“S.C.”) to measure each customer segment’s sensitivity to certain explanatory variables  
53 (e.g., weather, price, estimated efficiency improvements, and socioeconomic trends) that  
54 affect the segment’s natural gas usage.

55 **Q. What are North Shore’s current Service Classifications?**

56 A. North Shore’s customers are currently divided among three Service Classifications. The  
57 customers in S.C. No. 1 (Small Residential Service) and S.C. No. 2 (General Service) are  
58 classified as “firm general.” For test year 2015, North Shore forecasts 158,718 firm  
59 general customers. The 26 customers in S.C. No. 4 (Large Volume Demand Service) are  
60 classified as “large volume customers.”

61 **Q. How did North Shore determine its forecasted total demand?**

62 A. North Shore’s forecasted total demand is comprised of forecasts of its large volume  
63 customer demand and its firm general demand.

64 **Q. What methodology did North Shore use to determine the large volume customer  
65 demand forecast for the 2015 test year?**

66 A. There were a number of steps in the forecast process for large volume customer demand.  
67 First, monthly demands for S.C. Nos. 4 and above were summed<sup>1</sup>. Second, adjustments  
68 were made to the demand totals based on customer movement between Service  
69 Classifications. This ensured that the historical data and current large volume customer  
70 demand were comparable. Finally, a regression analysis was performed to forecast the  
71 long-term trend in large volume customer total demand.

72 **Q. What methodology did North Shore use to determine the firm general demand**  
73 **forecast for the 2015 test year?**

74 A. This analysis also had a number of steps. Demand was first divided into S.C. No. 1 and  
75 S.C. No. 2 demand. Each of these two classifications was further divided into demand by  
76 non-heating customers and demand by heating customers. Adjustments were then made  
77 to demand based on customer movement between Service Classifications in order to  
78 ensure that the historical data and current firm general customer demand were  
79 comparable. Finally, demand was divided into number of customers and usage per  
80 customer. This disaggregation of firm general demand provided the following eight  
81 components, which were forecasted independently on a monthly basis:

- 82 1) Usage per non-heating S.C. No. 1 customer
- 83 2) Number of non-heating S.C. No. 1 customers
- 84 3) Usage per heating S.C. No. 1 customer
- 85 4) Number of heating S.C. No. 1 customers
- 86 5) Usage per non-heating S.C. No. 2 customer
- 87 6) Number of non-heating S.C. No. 2 customers

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<sup>1</sup> S.C. No. 3 was re-numbered S.C. No. 4 in 2013.

- 88                   7)     Usage per heating S.C. No. 2 customer  
89                   8)     Number of heating S.C. No. 2 customers

90           The firm general demand was divided into these various components because various  
91           economic, demographic and weather factors affect each component of firm general  
92           demand differently. By examining each of the eight components, and relating them to  
93           those factors, a greater understanding is gained of how these factors affect firm general  
94           demand.

95   **Q.     Can you explain the S.C. No. 1 Heating forecast model in more detail?**

96   A.     Yes. The S.C. No. 1 Heating forecast uses two regression models, a number-of-  
97           customers model and a use-per-customer model. Both are monthly models and each was  
98           run with historical monthly data from January 2004 and January 2003 respectively to  
99           May 2013. The use-per-customer model is a regression using multiplicative variables  
100           developed by Itron<sup>2</sup> representing Heating and Other gas usage. Itron calls this a  
101           Statistically Adjusted End-Use (“SAE”) model. This model makes use of billing heating  
102           degree days (“HDD”), appliance saturation and efficiencies, home size (people per  
103           household), trends based on U.S. Energy Information Administration (“EIA”) data, real  
104           personal income, and real price to the customer.

105                   The number-of-customers model is based on the trend in the number of customers  
106           by S.C. and monthly binary variables. The total S.C. No. 1 Heating sales forecast is a  
107           combination of the use-per-customer model and the number-of-customer forecasts.

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<sup>2</sup> Itron is a technology provider to energy and water industries worldwide that developed both the multiplicative variables and the SAE regression models in the forecast. Itron provides measurement and control technology, communications systems, software and professional services to nearly 8,000 utilities in more than 100 countries. The Itron Forecasting Group develops and supports MetrixND – Itron’s statistical forecasting software tool used for short-term and long-term energy and demand forecasting. Itron’s MetrixND has more than 700 users from 170 utilities and energy companies around the world.

108 **Q. Please explain in more detail how the SAE models are used in the use-per-customer**  
109 **models.**

110 A. Using the S.C. No. 1 Heating forecast model as an example, the model design considers  
111 billing sales, price, structural changes, and appliance saturation and efficiencies trends. It  
112 then imposes a model structure through the SAE specification.

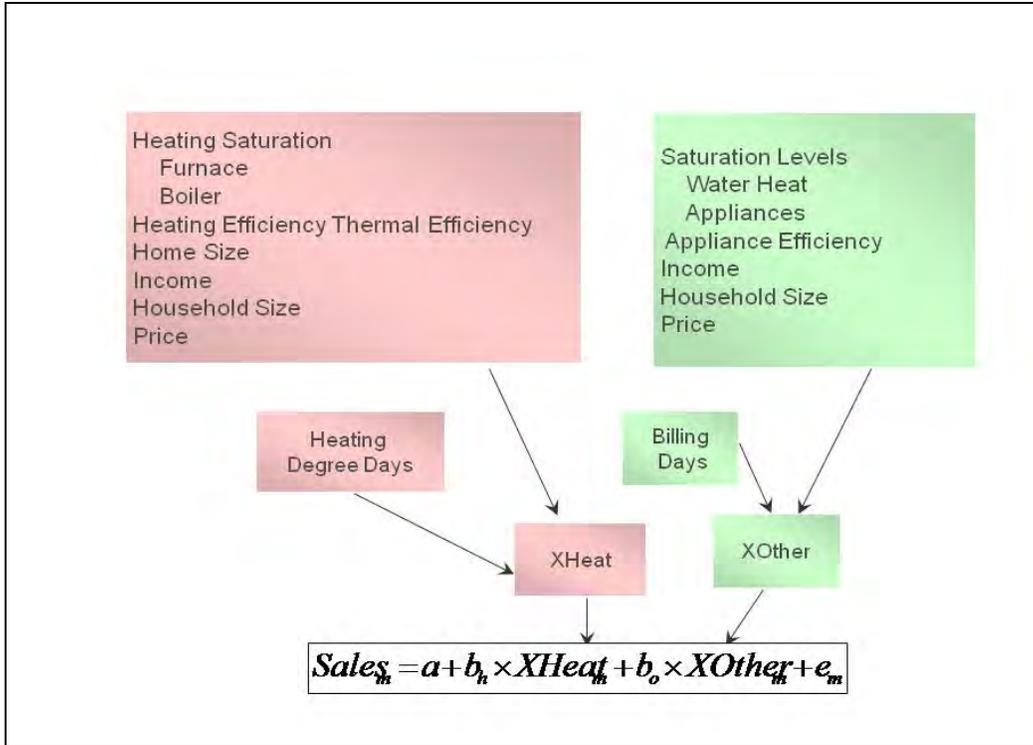
113 Instead of constructing a regression model with many explanatory variables, this  
114 approach constructs a model with two high-level end-use variables: Heating and Other  
115 Use. The model structure then embeds forecast drivers into these two constructed  
116 variables. The forecast drivers include HDD, price, income, household size (people per  
117 household), and end-use saturation and efficiency trends.

118 The estimated average use per customer regression model using the constructed  
119 end-use variables is:

$$120 \text{AvgUse}_t = B_0 + B_1X\text{Heat}_t + B_2X\text{Other}_t + e_t$$

121 The SAE model structure incorporates elasticity of demand, which is customers'  
122 behavior in response to changes in various explanatory variables, such as price, heating,  
123 cooling, income, etc. Customer behavior is based on research performed by Itron. By  
124 focusing on such customer behavior, North Shore can capture the appropriate impacts of  
125 changes in economic conditions and how they interrelate with end-use variables.

126 The graphic below explains in more detail the economic and various end-use  
127 saturation and efficiency variables, developed from the EIA energy efficiency forecasts,  
128 which make up the main explanatory variables:



129

130 The XHeat variable has two components:

131

$$XHeat_{y,m} = HeatIndex_y \times HeatUse_{y,m}$$

132

HeatIndex is expanded below:

133

$$HeatIndex_y = Structural Index_y \times \sum_{Type} Weight^{Type} \times \frac{\left( \frac{Sat_y^{Type}}{Eff_y^{Type}} \right)}{\left( \frac{Sat_{01}^{Type}}{Eff_{01}^{Type}} \right)}$$

134

HeatUse is expanded below:

135

$$HeatUse_{y,m} = \left( \frac{HDD_{y,m}}{HDD_{01}} \right) \times \left( \frac{HHSize_{y,m}}{HHSize_{01}} \right)^a \times \left( \frac{Income_{y,m}}{Income_{01}} \right)^b \times \left( \frac{Price_{y,m}}{Price_{01}} \right)^c$$

136

Factors impacting Heat Use or XHeat are:

137

1. Non-weather-sensitive end-use saturation and efficiency trends,

138

2. Number of heating degree days,

- 139                   3. Household size and income, and  
 140                   4. Prices.

141 **Q. Has North Shore used this model in the past to forecast firm general demand?**

142 A. Yes, North Shore has used this model in each of its last four rate cases.

143 **Q. How has the model performed historically?**

144 A. The model has performed well historically. The two types of equations – use per  
 145 customer and number of customers – have different characteristics and their statistical  
 146 reliability is quite high.

147 **B. Use-Per-Customer Equations**

148 **Q. How do you determine the statistical reliability of the use-per-customer equations?**

149 A. The statistical reliability of the use-per-customer equations is first measured with the  
 150 coefficient of determination, or  $R^2$ . The  $R^2$  measures the proportion or percentage of the  
 151 total variation in use per customer that is explained by the regression model. The  
 152 following table shows the  $R^2$  for each equation along with the percentage of actual  
 153 demand in 2012. Approximately 96 percent of the total variation in North Shore’s use  
 154 per customer is explained by the regression models.

<b>NS Use/Customer (S.C. Nos. 1-2) and Total Demand (S.C. No. 4)</b>		
	<b>Adjusted R-squared</b>	<b>Percentage of Demand</b>
<b>S.C. No. 1 Heating</b>	99.6%	53.5%
<b>S.C. No. 1 Non-heating</b>	90.1%	0.1%
<b>S.C. No. 2 Heating</b>	99.1%	26.5%
<b>S.C. No. 2 Non-heating</b>	86.0%	0.7%
<b>S.C. No. 4 (Total Demand)</b>	83.2%	19.2%
<b>Weighted Average</b>	96.2%	100.0%

155

156 **C. Number-of-Customers Equations**

157 **Q. How do you determine the statistical reliability of the number-of-customers**  
158 **equations?**

159 A. The statistical reliability of the number-of-customers equations is first measured with the  
160  $R^2$ . The following table shows the  $R^2$  for each equation along with the percentage of  
161 actual demand in 2012. Approximately 95 percent of the total variation in North Shore's  
162 number of customers is explained by the regression models.

<b>NS Number of Customers Equations</b>		
	<b>Adjusted R-squared</b>	<b>Percentage of Demand</b>
<b>S.C. No. 1 Heating</b>	96.4%	53.5%
<b>S.C. No. 1 Non-heating</b>	99.6%	0.1%
<b>S.C. No. 2 Heating</b>	98.4%	26.5%
<b>S.C. No. 2 Non-heating</b>	95.4%	0.7%
<b>S.C. No. 4</b>	85.2%	19.2%
<b>Weighted Average</b>	94.8%	100.0%

164 **Q. What general assumptions were made in developing the total demand forecast?**

165 A. The following assumptions were made:

- 166 • For S.C. Nos. 1 and 2 Heating, normal weather based on the twelve-year period 2001-  
167 2012 for Chicago, O'Hare weather station was used.<sup>3</sup> This equals 6,031 HDD for  
168 non-leap years and 6,054 HDD for leap years.
- 169 • Economic information is from the June 2013 Moody's Analytics forecast for North  
170 Shore.

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<sup>3</sup> The HDD using a 12 year normal was required by the Commission's Final Order in ICC Docket Nos. 07-0241/07-0242 (cons.). North Shore Gas Company used a 12-year normal in its three subsequent rate cases, and this has been uncontested.

- EIA forecasts of efficiency and saturation were provided by Itron.
- Price information was from NYMEX Short-Term Forecast dated June 13, 2013.

**Q. Based on these analyses, what level of customer demand does North Shore forecast for test year 2015?**

A. North Shore forecasts firm general demand of 27.3 Bcf and large volume customer demand of 6.7 Bcf, for a customer demand of 34.0 Bcf in test year 2015.

**Q. Was the demand forecast further allocated?**

A. Yes. The five classifications from the forecast of annual firm general demand volumes and customers were further divided into the following customer categories by month for volume blocking and revenue forecasting purposes. This approach determined monthly and annual volumes by Service Classification, revenue class (residential, commercial, and industrial), heating / non-heating, and sales type (retail and transport). The subgroups were:

Service Classification No. 1 Heating Forecast				Service Classification No. 1 Non-heating Forecast			
S.C. No. 1	Heating	Retail	Residential	S.C. No. 1	Non-heating	Retail	Residential
S.C. No. 1	Heating	Transport	Residential	S.C. No. 1	Non-heating	Transport	Residential
Service Classification No. 2 Heating Forecast				Service Classification No. 2 Non-heating Forecast			
S.C. No. 2	Heating	Retail	Residential	S.C. No. 2	Non-heating	Retail	Residential
S.C. No. 2	Heating	Transport	Residential	S.C. No. 2	Non-heating	Transport	Residential
S.C. No. 2	Heating	Retail	Commercial	S.C. No. 2	Non-heating	Retail	Commercial
S.C. No. 2	Heating	Transport	Commercial	S.C. No. 2	Non-heating	Transport	Commercial
S.C. No. 2	Heating	Retail	Industrial	S.C. No. 2	Non-heating	Retail	Industrial
S.C. No. 2	Heating	Transport	Industrial	S.C. No. 2	Non-heating	Transport	Industrial
Service Classification No. 4 Forecast							
S.C. No. 4	Heating	Transport	Industrial				

**Q. What was the basis of the allocation to the subgroups?**

186 A. The basis was historical sales by subgroup from 2012 actual sales volumes (“allocation  
187 base period”). The sales forecast was allocated based on the allocation base period  
188 percentages.

189 **Q. Were there any further allocations of the sales volume forecast?**

190 A. Yes. Sales volumes for S.C. Nos. 1 and 2 were allocated to the rate blocks (set amount or  
191 block of usage) using the monthly ogive curves (cumulative line graphs) developed from  
192 the billed frequency data for each of these customer classifications. These data were  
193 stored in the Revenue Forecasting Model (“RFM”), which blocked each month’s volumes  
194 individually for all S.C. No. 1 and S.C. No. 2 sub-groups by using the corresponding sub-  
195 group and months’ ogive curves from the allocation base period.

196 **III. COMPUTATION OF REVENUES BASED ON FORECAST**

197 **Q. Did North Shore use any other billing determinants besides volumes for revenue  
198 forecasting?**

199 A. Yes.

200 **Q. Please identify these other billing determinants and discuss how they were  
201 determined.**

202 A. The other billing determinants are as follows:

- 203 • Billing Periods: Base time period (January through December 2012) ratio of the  
204 number of billing periods to the number of customers × the forecasted number of  
205 customers. S.C. No. 2 billing periods were further allocated to small, medium and  
206 large meter classes based on the allocation of base time period’s monthly meter  
207 classes.

- 208 • Demand Volume: Analysis of two most recent months' demand volumes. S.C. No. 4  
209 only.
- 210 • Standby Demand Volume: Most recent month's standby demand volume.  
211 Transportation only.
- 212 • Standby Commodity Volume: Three year average of the monthly standby  
213 commodity volume percentage times transportation volume forecast. Transportation  
214 pool/contract only.
- 215 • Storage Gas Charge Volume: The storage gas charge monthly volume used in the  
216 forecast was calculated by looking at the most recent storage gas charge volume  
217 actuals.
- 218 • Storage Banking Charge Volume: Analysis of transportation customers' storage  
219 capacity.
- 220 • Demand Devices: Most recent month's demand device units. S.C. No. 2  
221 transportation only.
- 222 • Number of 2nd Pulse Units: Analysis of the two most recent months' 2nd pulse units.
- 223 • Number of Transportation Contract Accounts: Most recent month's number of  
224 transportation contract accounts. Transportation only.
- 225 • Number of Transportation Pool Accounts: Most recent month's number of  
226 transportation pool accounts and adjusted monthly for changes in number of  
227 transportation pool accounts. Transportation only.
- 228 • Number of Pools: Most recent month's number of pools. Transportation only.
- 229 • Number of Trades: Analysis of the base period trade count. Transportation  
230 pool/contract only.

- 231 • Number of Supplier Billing Option Credit Units: Most recent two months' average.
- 232 • Storage and Balancing Volume: Contract volume. Transportation pool/contract only.

233 **Q. What was done next?**

234 A. Revenues were calculated in the RFM.

235 **Q. How does the RFM calculate revenues?**

236 A. The RFM applies applicable rates to each billing determinants to calculate various  
237 revenues by month for all sub-groups. Specific revenue items, applicable billing  
238 determinants and rates are as follows:

- 239 • Customer Charge = number of billing periods × applicable customer charge rates.
- 240 • Demand Charge = demand volumes × demand rate.
- 241 • Storage Service Charge = retail sales volumes × storage service charge rate.
- 242 • Storage Banking Charge = transportation storage banking charge volume × storage  
243 banking charge rate.
- 244 • Demand Device Charge = number of demand devices × demand device rate.
- 245 • Distribution Charge = volumes in each block × applicable distribution charge rates.
- 246 • Volume Balancing Adjustment (Rider VBA) Charge = volumes × forecasted Rider  
247 VBA rates.
- 248 • Uncollectible Expense Adjustment Gas Cost (Rider UEA-GC) = retail gas charge  
249 revenue × Rider UEA-GC uncollectible factors.
- 250 • Franchise Cost Adjustment (Rider FCA) Charge = number of billing periods ×  
251 forecasted Rider FCA rate.
- 252 • 2nd Pulse Device Charge = number of 2nd pulse devices × 2nd pulse rate.

- 253 • Transportation Contract Administrative Charge = number of transportation contract  
254 accounts × transportation contract administrative charge rate.
- 255 • Transportation Pool Administrative Charge = number of transportation pool accounts  
256 × applicable transportation pool account administrative charge rates, plus number of  
257 transportation pools × transportation pool administrative charge rate.
- 258 • Transportation Trade Charge = number of trades × trade charge rate.
- 259 • Supplier Billing Option (Rider SBO) Credit = number of accounts forecast for  
260 supplier billing option × supplier billing option rate.
- 261 • Storage and Balancing Base Rate Revenue = storage and balancing volume × storage  
262 and balancing rates.
- 263 • Environmental Activities (Rider 11) Charge = volumes × forecasted Rider 11 rates.
- 264 • Uncollectible Expense Adjustment (Rider UEA) Charge = number of billing periods  
265 × forecasted Rider UEA rates.
- 266 • Energy Efficiency and On-bill Financing (Rider EOA) Charges = volumes ×  
267 forecasted Rider EOA rates.
- 268 • Renewable Energy Resources Fund Charge = number of billing periods × applicable  
269 rates.
- 270 • Low Income Energy Assistance Fund Charge = number of billing periods ×  
271 applicable rates.
- 272 • Retail Gas Charge Revenue = Retail gas sales volume × forecasted retail gas charge  
273 rates.
- 274 • Storage and Balancing Gas Charge = Storage and balancing volume × storage and  
275 balancing gas charge rates.

- 276 • Standby Demand Gas Charge Revenue = Standby demand volume × forecasted  
277 standby demand gas charge rates.
- 278 • Standby Commodity Gas Charge Revenue = Standby commodity volume ×  
279 forecasted standby commodity gas charge rates.
- 280 • Storage Gas Charge Revenue = Storage Gas Charge volume × forecasted Storage Gas  
281 Charge rates.
- 282 • Add-on Revenue Taxes = Taxable revenue × applicable add-on tax rates.
- 283 • Gas Use Taxes = Transportation volume × taxable therm percentage × applicable gas  
284 use tax rates.

285 **Q. How were the various rates determined?**

286 A. The North Shore tariff provided the rates for many of the base rate revenues and  
287 Regulatory Services provided the forecasted rates for various riders and gas charge  
288 revenues based on forecasted billing determinants and/or forecasted costs or revenues.

289 **IV. COMPARISON OF COMPARATIVE YEAR DEMAND AND FORECASTED**  
290 **DEMAND**

291 **Q. Please compare the 2015 test year demand forecast to the comparative year 2013**  
292 **(6 months of actual and 6 months forecast) demand.**

293 A. The comparative year 2013 demand is based on actual weather normalized demand  
294 (based on 6,031 HDD) January 2013 to June 2013 and forecasted demand for July  
295 through December 2013. The forecasted demand for 2013 is based on 6,126 normal  
296 heating degree days.

NSG Test Year Ending December 31, 2015 (Therms)								
Line No	Present Rate Classification	Fiscal Year 2013	Weather Adjustments	Normalized 2013	Test Year 2015	Difference	Annualized % Change	Line No
	(A)	(B)	(C)	(D)	(E)	(F)	(G)	
						(E-D)	(F/D)/2	
1	<b>Retail and Transportation</b>							1
2	Company Use	223,000	0	223,000	242,000	19,000	4.3%	2
3	S.C. No. 1	194,328,000	-6,168,000	188,160,000	179,537,000	-8,623,000	-2.3%	3
4	S.C. No. 2	98,743,000	-3,412,000	95,331,000	93,669,000	-1,662,000	-0.9%	4
5	S.C. No. 4	65,642,000	0	65,642,000	66,954,000	1,312,000	1.0%	5
6	<b>Total Volumes</b>	<b>358,936,000</b>	<b>-9,580,000</b>	<b>349,356,000</b>	<b>340,402,000</b>	<b>-8,954,000</b>	<b>-1.3%</b>	6

297

298 A. The declining annualized percent change from 2013 to 2015 for S.C. Nos. 1 and 2 is due  
 299 primarily to declining usage due to the energy efficiency gains documented by EIA  
 300 projections.

301 **Q. Does this conclude your direct testimony?**

302 A. Yes, it does.