

“Top Down” Estimation of DSM Program Impacts on Natural Gas Usage: Summary of Findings

Larry Kaufmann, *Senior Advisor*

Pacific Economics Group

May 13, 2010

Toronto, Ontario



Introduction

Some parties in Ontario have raised concerns about the “bottom up” approach that is currently used to estimate savings from utility DSM programs

- Depends on a number of engineering assumptions
- Leads to controversial and costly regulation

It has been suggested that instead of starting with individual DSM programs that a “top down” approach using econometric methods to estimate savings from utility DSM programs could be used

Summary

Pacific Economics Group Research (PEG) was asked to explore whether “top down” estimates of DSM savings could be developed using available data in Ontario

Using available data in Ontario PEG explored many options but was not able to develop a “top down” model that can be used to estimate savings from utility DSM programs

However, we did find some indicative evidence of the impact of DSM spending on gas consumption

i.e. 10% increase DSM expenditures

→ 0.56% to 1.05% decline in residential gas consumption

→ 0.34% to 0.89% decline in commercial gas consumption

Outline

1. “Top Down” Precedents and Experience in the Industry
2. Existing Gas Demand Models in Ontario
3. PEG’s Top Down Gas Demand Models and Econometric Results
4. Conclusion and Next Steps

I. Precedents and Industry Experience

Jurisdictional survey did not find any examples of the “top down” approach proposed in Ontario

However, econometric research on energy conservation and utility DSM programs is extensive

A noteworthy article is “Demand Side Management and Energy Efficiency in the United States,” by David Loughran and Jonathan Kulick

- used an econometric specification that examines the relationship between *changes* in electricity consumption and changes in independent variables, including DSM spending
- findings indicated that DSM expenditures reduce electricity sales by 0.4% to 0.6% per annum

II. Existing Gas Demand Models in Ontario

Enbridge Gas Distribution (EGD) and Union Gas both currently use econometric models to forecast gas usage in regulatory proceedings

PEG's review found that, while the EGD and Union models may be appropriate for forecasting, they were less suited for developing "top down" estimates of gas savings from their DSM programs

Top down models require statistically robust estimates of the impact of gas DSM programs on gas consumption

The EGD and Union models had technical imperfections (serial correlation, heteroskedasticity) that could bias statistical inference on whether gas DSM programs have a significant effect on gas consumption

These imperfections are less problematic for their current forecasting applications

III. PEG's Top Down Models

PEG investigated three approaches to top-down estimation of gas savings

- A. New gas demand models
- B. Replicate variants of company demand models but include DSM spending as an independent variable
- C. Examine the difference between actual gas consumption and gas consumption predicted by gas demand models that exclude DSM as an explanatory variable; a statistically significant difference could be an indicator of the impact of DSM on gas consumption

A. New Gas Demand Models

PEG used a two-stage approach to modeling gas demand

First Stage

- Regress monthly values of gas consumption on heating degree days and final delivered prices for natural gas
- Use these results to develop normalized gas consumption volumes

Separate Regressions for:

	<u>Enbridge</u>	<u>Union</u>
Residential	Revenue Class 20	Revenue Class 01-R Revenue Class M2-R
Commercial	Revenue Class 12 Revenue Class 48	Revenue Class 01-C Revenue Class M2-C Revenue Class 10



A. New Gas Demand Models (Con't)

Second Stage

- Aggregate (normalized) monthly consumption values into annual values
- Results use annual data, since data on some variables (including DSM spending) only available annually
- Regress changes in normalized gas consumption against a variety of other economic and customer characteristic variables that can impact gas consumption
- One variable is utility gas DSM spending in previous year
- Coefficient on DSM variable can represent impact of DSM spending on next year's gas consumption

Both first and second stage regressions also correct for technical imperfections recognized in EGD and Union models

A. New Gas Demand Models (Con't)

Second Stage

Regressions done for residential and commercial customers of both companies

- Residential customers for EGD and Union residential revenue classes stacked in same regression; same for commercial customers
- Necessary to stack revenue classes to have large enough samples to obtain statistically significant results
- Separate constant terms in this regression for EGD and Union to control for differences in EGD and Union company characteristics
- Nevertheless, relatively small sample size only allows estimates of how DSM expenditures impact gas savings for residential customers in Ontario, and commercial customers in Ontario, NOT different estimates of savings for EGD or Union

A. New Gas Demand Models (Con't)

PEG's **first stage** regressions generally sensible

For all eight revenue classes

- Heating degree days positively related to gas consumption
- Price negatively related to gas consumption
- Highly statistically significant estimates

Regressions also had high R-squared statistics (from 0.68 to 0.92), which measures the percent of change in the dependent variable that is accounted for by change in the independent variables

A. New Gas Demand Models (Con't)

PEG explored dozens of **second stage** specifications, with different choices for customer and economic characteristics that impact consumption

Best results for residential customers showed the following explanatory variables had the expected sign and a statistically significant impact on consumption

- Unemployment rate in Ontario (negative impact on consumption)
- Number of customers per household (positive impact)
- Customer 'vintage' (negative impact from more recent customers)
- Number of EcoEnergy evaluations (negative impact on consumption)

However, no regression showed that utility DSM spending had an impact on residential customers' gas consumption



A. New Gas Demand Models (Con't)

Best results for **second stage** regressions for commercial customers even less satisfactory

Only significant driver of commercial gas consumption was the commercial vacancy rate (negative impact)

No regression showed that utility DSM spending had a negative impact on commercial customers' gas consumption (see Appendix A for results)

B. Update Variants of Company Gas Demand Models to Include DSM Spending

PEG also updated variants of the gas demand models the companies use, but added DSM spending as an independent variable

PEG's models corrected for statistical imperfections in company estimates

This approach required estimates of monthly DSM spending

Gas DSM data originally not collected monthly, so these data are less reliable than annual data

Relationship between monthly gas DSM spending and savings also less reliable due to:

- quality of DSM data
- varying relationship during year between expenditures and resulting changes in consumption



B. Update Variants of Company Gas Demand Models to Include DSM Spending (Con't)

PEG's specifications using monthly estimates of DSM spending, to increase number of sample observations, had more success (see Appendix B for results)

DSM has a negative and significant impact on gas consumption for all three residential customer classes and for two of five commercial classes

- For Enbridge: a 10% increase in DSM expenditures will be associated with contemporaneous declines in gas consumption of 1.05% for Residential revenue class 20 and 0.84% for Commercial revenue class 12.
- For Union: a 10% increase in DSM expenditures will be associated with contemporaneous declines in gas consumption of 0.77% for Residential 01 customers, 0.56% for Residential M2 customers, and 0.34% for Commercial 01 customers.

C. Comparisons Between Actual and Predicted Demand

PEG also compared whether there are statistically significant differences between actual and predicted changes in gas consumption for both annual and monthly regressions

Predicted gas demand based on gas demand models that exclude DSM as an explanatory variable

Models also generate confidence intervals around predictions

If actual gas consumption is less than the predicted value and outside of the confidence interval, gas consumption is significantly lower than what's predicted by 'conventional' gas demand model

>>>> this difference could be an indicator of the impact of DSM programs on gas consumption

C. Comparisons Between Actual and Predicted Demand (Con't)

PEG's final approach also never showed any instances where actual gas consumption was less than the predicted value and outside of the confidence interval

This approach therefore does not provide any indicative, econometric evidence of gas savings from DSM spending

V. Conclusions

PEG's research did not identify any statistical evidence linking utility DSM spending and subsequent declines in gas consumption that is solid enough to develop “top down” estimates of gas DSM savings

As noted, we are not aware of any DSM plans in the industry that currently use such “top down” estimates

However, we did find some indicative evidence of the impact of DSM spending on gas consumption

Residential customers: 10% DSM Spending → 0.6% to 1% decline

Commercial customers: 10% DSM Spending → 0.3% to 0.8% decline



V. Conclusions(Con't)

PEG analysis could be improved if better data were available

- Monthly DSM spending by rate class and zone
- Information on when measures were actually installed (and not just \$ spent)

More appropriate estimates may also be developed if demand models are estimated separately for participating and non-participating customers (as in California)

However, developing detailed customer-specific data:

- Is more costly
- Would take years to develop sufficient samples
- Would not necessarily yield significant results

Appendix A: Econometric Results

Table 1

First Stage Regression: Average Gas Use Per Customer Enbridge Revenue Class 20

VARIABLE KEY

HDD= Heating Degree Days for Revenue Class 20

P= Residential Total Delivery Price for Revenue Class 20

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
HDD	0.288	189.775	P	-0.129	-2.826
Constant	5.672	453.120	Trend	-0.005	-2.512

Other Results

System Rbar-Squared	0.721
Durbin-Watson Statistic	1.783
F Statistic	139.976
Sample Period	1991-2008
Number of Observations	216



Appendix A: Econometric Results

Table 2

First Stage Regression: Average Gas Use Per Customer Enbridge Revenue Class 12

VARIABLE KEY

HDD= Heating Degree Days for Revenue Class 12

P= Commercial Total Delivery Price for Revenue Class 12

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
HDD	0.278	143.320	P	-0.263	-6.947
Constant	8.842	894.565	Trend	0.031	15.107

Other Results

System Rbar-Squared	0.686
Durbin-Watson Statistic	2.077
F Statistic	118.41
Sample Period	1991-2008
Number of Observations	216



Appendix A: Econometric Results

Table 3

First Stage Regression: Average Gas Use Per Customer Enbridge Revenue Class 48

VARIABLE KEY

HDD= Heating Degree Days for Revenue Class 48

P= Commercial Total Delivery Price for Revenue Class 48

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
HDD	0.344	355.379	P	-0.055	-2.812
Constant	7.324	889.740	Trend	0.012	11.511

Other Results

System Rbar-Squared	0.753
Durbin-Watson Statistic	1.818
F Statistic	164.96
Sample Period	1991-2008
Number of Observations	216



Appendix A: Econometric Results

Table 4

First Stage Regression: Average Gas Use Per Customer Union Revenue Class 01 Residential

VARIABLE KEY

HDD= Heating Degree Days for Revenue Class 01 Residential

P= Total Delivery Price for Revenue Class 01 Residential

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
HDD	0.556	79.588	P	-0.629	-17.665
Constant	5.385	362.058	Trend	0.010	6.099

Other Results

System Rbar-Squared	0.917
Durbin-Watson Statistic	1.764
F Statistic	591.97
Sample Period	1991-2008
Number of Observations	216



Appendix A: Econometric Results

Table 5

First Stage Regression: Average Gas Use Per Customer Union Revenue Class 01 Commercial

VARIABLE KEY

HDD= Heating Degree Days for Revenue Class 01 Commercial

P= Total Delivery Price for Revenue Class 01 Commercial

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
HDD	0.683	82.750	P	-0.604	-21.069
Constant	6.561	831.924	Trend	0.012	9.109

Other Results

System Rbar-Squared	0.881
Durbin-Watson Statistic	1.629
F Statistic	398.48
Sample Period	1991-2008
Number of Observations	216



Appendix A: Econometric Results

Table 6

First Stage Regression: Average Gas Use Per Customer Union Revenue Class M2 Residential

VARIABLE KEY

HDD= Heating Degree Days for Revenue Class M2 Residential

P= Total Delivery Price for Revenue Class M2 Residential

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
HDD	0.296	87.312	P	-1.325	-48.956
Constant	5.138	408.929	Trend	0.037	25.273

Other Results

System Rbar-Squared	0.822
Durbin-Watson Statistic	1.892
F Statistic	248.52
Sample Period	1991-2008
Number of Observations	216



Appendix A: Econometric Results

Table 7

First Stage Regression: Average Gas Use Per Customer Union Revenue Class M2 Commercial

VARIABLE KEY

HDD= Heating Degree Days for Revenue Class M2 Commercial

P= Total Delivery Price for Revenue Class M2 Commercial

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
HDD	0.406	169.847	P	-0.058	-58.389
Constant	7.240	1419.319	Trend	0.016	17.251

Other Results

System Rbar-Squared	0.802
Durbin-Watson Statistic	1.575
F Statistic	218.74
Sample Period	1991-2008
Number of Observations	216



Appendix A: Econometric Results

Table 8

First Stage Regression: Average Gas Use Per Customer Union Revenue Class 10

VARIABLE KEY

HDD= Heating Degree Days for Revenue Class 10

P= Total Delivery Price for Revenue Class 10

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
HDD	0.595	252.715	P	-0.248	-22.629

Constant 9.035 1383.267

Other Results

System Rbar-Squared 0.873

Durbin-Watson Statistic 1.647

F Statistic 370.45

Sample Period 1991-2008

Number of Observations 216



Appendix A: Econometric Results

Table 9

Second Stage Regression: Change in 'Normalized' Gas Use Residential Revenue Classes

VARIABLE KEY

ID1= Constant for Enbridge Revenue Class 20
ID2= Constant for Union Revenue Class 01 Residential
ID3= Constant for Union Revenue Class M2 Residential
ECO= EcoEnergy dummy variable
DSM= DSM Spending in previous year
UR= Unemployment Rate
VIN= Customer Vintage
NPHH= Number of People per Household

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
ID1	-2.641	-2.545	ECO	-0.047	-4.315
ID2	-2.494	-2.509	DSM	-0.009	-0.496
ID3	-2.576	-2.510	UR	-0.005	-3.223
			VIN	0.259	2.026
			NPHH	2.385	2.550

Other Results

System Rbar-Squared	0.388
Durbin-Watson Statistic	2.86
F Statistic	3.18
Sample Period	1999-2008
Number of Observations	30



Appendix A: Econometric Results

Table 10

Second Stage Regression: Change in 'Normalized' Gas Use Commercial Revenue Classes

VARIABLE KEY

ID1= Constant for Enbridge Rate Class 6
ID2= Constant for Union Revenue Class 01 Commercial
ID3= Constant for Union Revenue Class M2 Commercial
ID4= Constant for Union Revenue Class 10
BUC= 2005, 2006 dummy variables
DSM= DSM Spending in previous year
SEGM= Segmentation index
CVR= Commercial Vacancy Rate

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
ID1	-0.402	-1.117	BUC	-0.001	-0.188
ID2	-0.315	-0.831	DSM	0.017	1.320
ID3	-0.331	-0.877	SEGM	0.343	0.900
ID4	-0.083	-0.876	CVR	-0.006	-3.307

Other Results

System Rbar-Squared	0.188
Durbin-Watson Statistic	2.354
F Statistic	2.00
Sample Period	1999-2008
Number of Observations	40



Appendix B: Econometric Results

Table 11

Alternate Regression: Monthly DSM Data Enbridge Revenue Class 20

VARIABLE KEY

P= Total Delivery Price for Revenue Class 20
HDD= Heating Degree Days for Revenue Class 20
FE= Furnace Efficiency Index
ECOE= Eco Energy dummy variable
DSM= DSM Cost for Rate Class 1

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
P	-0.172	-1.596	ECOE	-0.023	-5.266
HDD	0.335	34.769	DSM	-0.105	-18.717
FE	-0.328	-11.045			
Constant	5.691	74.654	Trend	0.027	4.112

Other Results

System Rbar-Squared	0.735
Durbin-Watson Statistic	1.811
F Statistic	52.69
Sample Period	1998-2008
Number of Observations	132



Appendix B: Econometric Results

Table 12

Alternate Regression: Monthly DSM Data Enbridge Revenue Class 12

VARIABLE KEY

P= Total Delivery Price for Revenue Class 12
HDD= Heating Degree Days for Revenue Class 12
BUC1= Building Code 2005 dummy variable
BUC2= Building Code 2006 dummy variable
RM= Rate Migration dummy variable
DSM= DSM Cost for Rate Class 6

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
P	-0.263	-27.714	BUC2	-0.015	-10.238
HDD	0.251	134.269	RM	-0.005	-5.012
BUC1	-0.015	-4.370	DSM	-0.084	-5.232
Constant	8.237	174.744	Trend	0.083	18.350

Other Results

System Rbar-Squared	0.68
Durbin-Watson Statistic	1.792
F Statistic	35.69
Sample Period	1998-2008
Number of Observations	132

Appendix B: Econometric Results

Table 13

Alternate Regression: Monthly DSM Data Enbridge Revenue Class 48

VARIABLE KEY

P= Total Delivery Price for Revenue Class 48
HDD= Heating Degree Days for Revenue Class 48
ONTGDP= Ontario GDP
CVR= GTA Commercial Vacancy Rate
DSM= DSM Cost for Rate Class 6
Trend= Time Trend

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
P	-0.684	-4.342	CVR	0.137	1.815
HDD	0.301	47.370	DSM	-0.021	-1.176
ONTGDP	2.266	4.250	Trend	0.003	0.169
Constant	7.328	45.301			

Other Results

System Rbar-Squared	0.734
Durbin-Watson Statistic	1.702
F Statistic	52.63
Sample Period	1998-2008
Number of Observations	132



Appendix B: Econometric Results

Table 14

Alternate Regression: Monthly DSM Data Union Revenue Class 01 Residential

VARIABLE KEY

P= Total Delivery Price for Revenue Class 01 Residential
HDD1= January Heating Degree Days for Northern Region
HDD2= February Heating Degree Days for Northern Region
HDD3= March Heating Degree Days for Northern Region
HDD4= April Heating Degree Days for Northern Region
HDD5= May Heating Degree Days for Northern Region
HDD9= September Heating Degree Days for Northern Region
HDD10= October Heating Degree Days for Northern Region
HDD11= November Heating Degree Days for Northern Region
HDD12= December Heating Degree Days for Northern Region
NPHH= Number of Persons Per Household
FE= Furnace Efficiency Index
ECOE= Eco Energy dummy variable
DSM= DSM Cost for Rate 01 Residential

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
P	-0.339	-16.540	HDD10	0.050	22.368
HDD1	0.131	50.066	HDD11	0.102	42.915
HDD2	0.130	54.414	HDD12	0.136	60.714
HDD3	0.117	49.760	NPHH	-0.107	-0.885
HDD4	0.080	38.278	FE	0.155	7.748
HDD5	0.038	16.079	ECOE	0.001	0.900
HDD9	0.010	3.217	DSM	-0.077	-10.367
Constant	4.124	1563.138			

Other Results

System Rbar-Squared	0.973
Durbin-Watson Statistic	1.893
F Statistic	318.58
Sample Period	1998-2008
Number of Observations	132



Appendix B: Econometric Results

Table 15

Alternate Regression: Monthly DSM Data Union Revenue Class M2 Residential

VARIABLE KEY

P= Total Delivery Price for Revenue Class M2 Residential
HDD1= January Heating Degree Days for Southern Region
HDD2= February Heating Degree Days for Southern Region
HDD3= March Heating Degree Days for Southern Region
HDD4= April Heating Degree Days for Southern Region
HDD5= May Heating Degree Days for Southern Region
HDD9= September Heating Degree Days for Southern Region
HDD10= October Heating Degree Days for Southern Region
HDD11= November Heating Degree Days for Southern Region
HDD12= December Heating Degree Days for Southern Region
NPHH= Number of Persons Per Household
FE= Furnace Efficiency Index
ECOE= Eco Energy dummy variable
DSM= DSM Cost for Rate M2 Residential

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
P	-0.214	-10.319	HDD10	0.053	29.520
HDD1	0.137	43.947	HDD11	0.096	61.445
HDD2	0.138	81.708	HDD12	0.142	109.293
HDD3	0.126	69.459	NPHH	-0.038	-0.204
HDD4	0.084	53.539	FE	0.112	9.645
HDD5	0.037	28.495	ECOE	-0.001	-1.515
HDD9	-0.007	-4.970	DSM	-0.056	-8.587
Constant	4.129	773.148			

Other Results

System Rbar-Squared	0.984
Durbin-Watson Statistic	1.684
F Statistic	544.56
Sample Period	1998-2008
Number of Observations	132



Appendix B: Econometric Results

Table 16

Alternate Regression: Monthly DSM Data Union Revenue Class 01 Commercial

VARIABLE KEY

P= Total Delivery Price for Revenue Class 01 Commercial
HDD1= January Heating Degree Days for Northern Region
HDD2= February Heating Degree Days for Northern Region
HDD3= March Heating Degree Days for Northern Region
HDD4= April Heating Degree Days for Northern Region
HDD5= May Heating Degree Days for Northern Region
HDD9= September Heating Degree Days for Northern Region
HDD10= October Heating Degree Days for Northern Region
HDD11= November Heating Degree Days for Northern Region
HDD12= December Heating Degree Days for Northern Region
SEGM= Segmentation Index
DSM= DSM Cost for Rate 01 Commercial

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
P	-0.125	-3.948	HDD9	0.009	5.950
HDD1	0.181	91.066	HDD10	0.095	71.405
HDD2	0.175	145.277	HDD11	0.135	79.792
HDD3	0.162	274.763	HDD12	0.174	131.892
HDD4	0.114	129.495	SEGM	0.164	0.949
HDD5	0.060	78.135	DSM	-0.034	-3.825
Constant	5.100	736.159			

Other Results

System Rbar-Squared	0.956
Durbin-Watson Statistic	1.991
F Statistic	221.43
Sample Period	1998-2008
Number of Observations	132



Appendix B: Econometric Results

Table 17

Alternate Regression: Monthly DSM Data Union Revenue Class M2 Commercial

VARIABLE KEY

P= Total Delivery Price for Revenue Class M2 Commercial
HDD1= January Heating Degree Days for Southern Region
HDD2= February Heating Degree Days for Southern Region
HDD3= March Heating Degree Days for Southern Region
HDD4= April Heating Degree Days for Southern Region
HDD5= May Heating Degree Days for Southern Region
HDD9= September Heating Degree Days for Southern Region
HDD10= October Heating Degree Days for Southern Region
HDD11= November Heating Degree Days for Southern Region
HDD12= December Heating Degree Days for Southern Region
SEGM= Segmentation Index
DSM= DSM Cost for Rate M2 Commercial

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
P	-0.128	-6.992	HDD9	0.014	22.459
HDD1	0.163	114.384	HDD10	0.082	219.577
HDD2	0.160	102.119	HDD11	0.129	113.005
HDD3	0.150	221.101	HDD12	0.158	198.070
HDD4	0.106	98.564	SEGM	0.515	4.625
HDD5	0.060	143.988	DSM	-0.002	-0.268
Constant	5.949	1229.972			

Other Results

System Rbar-Squared	0.968
Durbin-Watson Statistic	1.891
F Statistic	309.49
Sample Period	1998-2008
Number of Observations	132



Appendix B: Econometric Results

Table 18

Alternate Regression: Monthly DSM Data Union Revenue Class 10 Commercial

VARIABLE KEY

P= Total Delivery Price for Revenue Class 10 Commercial
HDD1= January Heating Degree Days for Northern Region
HDD2= February Heating Degree Days for Northern Region
HDD3= March Heating Degree Days for Northern Region
HDD4= April Heating Degree Days for Northern Region
HDD5= May Heating Degree Days for Northern Region
HDD9= September Heating Degree Days for Northern Region
HDD10= October Heating Degree Days for Northern Region
HDD11= November Heating Degree Days for Northern Region
HDD12= December Heating Degree Days for Northern Region
SEGM= Segmentation Index
DSM= DSM Cost for Rate 10 Commercial

EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC	EXPLANATORY VARIABLE	PARAMETER ESTIMATE	T-STATISTIC
P	-0.070	-3.425	HDD9	-0.002	-0.504
HDD1	0.152	124.821	HDD10	0.083	105.458
HDD2	0.140	453.624	HDD11	0.117	458.747
HDD3	0.137	175.920	HDD12	0.136	319.824
HDD4	0.090	57.349	SEGM	-0.208	-2.139
HDD5	0.043	61.816	DSM	-0.003	-0.467
Constant	7.842	3622.985			

Other Results

System Rbar-Squared	0.94
Durbin-Watson Statistic	1.825
F Statistic	157.93
Sample Period	1998-2008
Number of Observations	132

