

**Exhibit No. VN-4
Dockets UE-090704 and UG-090705
Witness: Vanda Novak**

**BEFORE THE WASHINGTON STATE
UTILITIES AND TRANSPORTATION COMMISSION**

**WASHINGTON UTILITIES AND
TRANSPORTATION COMMISSION,**

Complainant,

v.

PUGET SOUND ENERGY, INC.,

Respondent.

DOCKET UE-090704

DOCKET UG-090705

EXHIBIT TO TESTIMONY OF

VANDA NOVAK

**STAFF OF
WASHINGTON UTILITIES AND
TRANSPORTATION COMMISSION**

Company Response to Staff Data Request No. 188

November 17, 2009

BEFORE THE WASHINGTON UTILITIES AND TRANSPORTATION COMMISSION

Docket Nos. UE-090704 and UG-090705
Puget Sound Energy, Inc.'s
2009 General Rate Case

WUTC STAFF DATA REQUEST NO. 188

WUTC STAFF DATA REQUEST NO. 188:

Re: Weather Normalization

On pages 12-13 of Ms. Molander's direct testimony, Exhibit (LIM-1T), she explains a modification to the gas rate schedule equations in the 2006 weather normalization methodology, which "allows the estimated seasonal coefficient to be more robust by including more observations per seasonal coefficient." (Page 13, Lines 3-4.) Please provide a representative sample of the analysis the Company conducted to reach this conclusion and a written narrative describing the analysis.

Response:

Attached as Attachment A to Puget Sound Energy, Inc.'s ("PSE") Response to WUTC Staff Data Request No. 188, please find an MS Excel file that contains regression results and statistics supporting the modification of the individual monthly weather variables to seasonal weather variables in the gas rate schedule model. Regression results and statistics for residential Schedule 23 and commercial Schedule 31 are provided.

In the weather normalization methodology approved in PSE's general rate case ("GRC"), WUTC Docket Nos. UE-060266 and UG-060267, the gas rate schedule equations allowed the estimated coefficients on weather variables to vary uniquely by each calendar month. PSE modified this methodology for this proceeding by restricting the weather sensitivity coefficients to be identical across months in the same season, such as a winter coefficient and a shoulder season coefficient. The benefit of restricting the weather coefficients to be equal across a set of months is to increase the number of observations per coefficient that the gas rate schedule model has to estimate the relationship between weather and load. Increasing the number of observations per coefficient in this way is particularly useful for the gas rate schedule model because it is based on monthly data.

Support for the use of restricting the weather coefficients to be equal across seasons versus using unique monthly weather variables rests on the following set of statistical assumptions: 1) the coefficients on the seasonal weather variables are statistically

significant or if a coefficient is insignificant, it is more significant than the unique monthly coefficients it has replaced, 2) the seasons display the expected sign and relationships between season. For example, winter coefficients would be expected to be positive, and greater than shoulder season coefficients.

Please see the tab 'Schedule 23' in Attachment A to PSE's Response to WUTC Staff Data Request No. 188 to compare the 2006 GRC methodology and the revised methodology. Using the 2006 GRC methodology updated with load and temperature information for this proceeding, the model has estimated inconsistent coefficients on the HDD variables (in column H). For example, the coefficient on January HDDs, 0.117, is statistically significant and has the expected sign (positive), implying that colder weather leads to greater load. However, the coefficient on February, -0.09385, is statistically insignificant and has a counter-intuitive negative sign.

Using the revised methodology, it can be seen that the winter coefficient (WIN2HDD) has the expected sign (positive), and is highly significant. The "Std. Error" column reports the estimated standard errors of the coefficient estimates. The standard errors measure the statistical reliability of the coefficient estimates—the larger the standard errors, the more statistical noise are in the estimates. The WIN2HDD variable has lower standard error, 0.0098, than any of the monthly weather coefficients in the 2006 GRC model (column I). Although the SHLD2HDD variable does not pass the 5% significance test, it has a lower standard error than the two variables it replaced, the June and September HDD coefficients in the 2006 GRC model, implying a better estimate of fit. Similar results are shown for the Schedule 31 commercial model.

ATTACHMENT A to PSE's Response to WUTC Staff Data Request No. 188

Gas Schedule 23 Load (User-per-Customer) Equation

Revised Methodology (Using Seasonal Coefficients)

Dependent Variable: S23RU
 Method: Least Squares
 Date: 10/20/09 Time: 16:23
 Sample: 2004/01/208B/09
 Included observations: 57
 Convergence achieved after 10 iterations

Variable	Coefficient	Std. Error	t-Statistic	Prob.	Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	19.86414	1.142772	17.38242	0.0000	C	78.21211	25.91127	3.01846	0.00448
JAN	37.41374	7.514054	4.979168	0.0000	JAN	30.33031	1.2384	0.2241	
FEB	24.57903	6.375083	3.856584	0.0004	FEB	38.87253	1.730058	0.0927	
MAR	15.86746	6.056498	2.556031	0.0132	MAR	-22.2044	27.86693	-0.7988	
APR	0.785559	4.87174	0.161248	0.8727	APR	-63.8412	27.11451	-2.3545	0.4311
MAY	-8.46814	3.281137	-1.97131	0.0551	MAY	-79.68981	26.68893	-2.68444	0.0115
JUN	2.43282	4.165132	0.584092	0.5622	JUN	-55.4483	26.14737	-2.12061	0.0413
SEP	4.454915	4.240582	1.050453	0.2993	SEP	25.9502	1.24813	0.0312	
OCT	1.236998	4.384733	0.282115	0.7792	OCT	53.3841	25.94881	2.28915	0.0310
NOV	19.53984	6.519624	2.89708	0.0045	NOV	-5.83406	28.81253	-0.193345	0.9516
DEC	35.87517	7.493119	4.787748	0.0000	DEC	-42.8347	28.43433	-1.50986	0.1403
WIN2HDD	0.089745	0.09877	10.98978	0.0000	WIN2HDD	-62.8576	43.7078	-1.42705	0.1998
SHLD2HDD	0.030989	0.022694	1.355985	0.1791	SHLD2HDD	0.1117	0.012886	8.65904	0.0000
AR(1)	0.32039	0.139324	2.289595	0.0264	AR(1)	-0.09385	0.066362	-1.41415	0.1664
R-squared	0.995976	Mean dependent v 67.68893			MARHDD	0.064535	0.019453	3.220803	0.0022
Adjusted R-s	0.984759	S.D. dependent v 41.1512			APRHDD	0.113171	0.018329	6.010546	0.0000
S.E. or regre	2.979105	Akaike info criterion 5.2305			MAYHDD	0.152273	0.052492	2.900899	0.0065
Sum squared	381.6278	Schwarz criterion 5.7323			JUNHDD	0.021813	0.023201	1.211698	0.2340
Log likelihood	-135.069	Hannan-Quinn crit 5.4255			SEPHDD	0.019126	0.037809	0.605459	0.6144
F-statistic	818.6285	Durbin-Watson stat 2.0093			OCTHDD	0.064355	0.030576	2.104732	0.0428
Prob(F-statist	4.92E-47				NOVHDD	0.1378	0.051411	2.655687	0.0118
Inverted AR	0.32				DECHDD	0.063052	0.035289	1.956735	0.0586
					AR(1)	0.522626	0.151027	3.460488	0.0015

Inverted AR

0.32

R-squared 0.997276
 Adjusted R-squared 0.995513
 S.D. dependent v 41.1512
 S.E. of reg 2.756525
 Akaike info criterion 5.156145
 Sum squared 258.3466
 Schwarz criterion 5.380534
 Log likelihood -123.95
 Hannan-Quinn crit 5.476531
 F-statistic 565.7463
 Prob(F-st 2.08E-37

Inverted A

0.52

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Equation Variables Key

S23RU = Schedule 23 (residential) User-per-Customer
 JAN, FEB, ..., NOV = Monthly Dummy Variables
 JANHDD, FEBHDD, ..., DECHDD = Monthly Total of Heating degree Days with a 65 F cut point
 WIN2HDD = Winter Seasonal Dummy (Oct, Nov, Dec, Jan, Feb, Mar, Apr, May)* [Monthly Total of Heating Degree Days with a 65 F cut point]
 SHLD2HDD = Shoulder Seasonal Dummy (June, September)* [Monthly Total of Heating Degree Days with a 65 F cut point]

