

AN EVALUATION OF NATURAL GAS CONSERVATION IN
THE RESIDENTIAL SECTOR OF OHIO
USING TIME-VARYING-PARAMETER MODELS

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EXECUTIVE SUMMARY

This paper estimates and evaluates conservation of natural gas in the residential sectors of the four largest distribution utilities in Ohio. The case studies include monthly data over the period from January 1970 until March 1980, and therefore, are comprehensive in the sense of including data on periods before, during, and after conservation pressures and incentives.

Both conventional and new time-varying-parameter models are combined in this paper in order to handle the dynamic behavior of consumers and to provide new information to the analyst in model construction. By themselves, conventional energy demand models in linear and log-linear forms fail to produce useful results due to the fact that base-price and personal-disposable-income effects are collinear and therefore not separable. As a result, estimated coefficients for conventional models have incorrect signs and tend to be statistically insignificant. Instead of the linear or log-linear forms, time-varying-parameter models show that the ratio of income and price, interacted with heating degree days, is powerful in explaining consumer behavior. Thus, the ratio form is used in this paper.

Average price data, obtained by dividing total residential revenue by total sales, are found to be an inappropriate variable for natural gas modeling. The fixed cost component of total price drives up the summertime average price, giving the spurious impression that high summer prices result in low consumption levels. This is incorrect as, of course, warm weather eliminates the use of natural gas for space heating in summer, and this is the cause for low consumption levels in summer. The marginal price variable does not suffer from this problem and is desirable for modeling consumer behavior but is difficult to obtain. To supply the needed marginal price data, this paper introduces a new time-varying-parameter model that estimates marginal prices from the readily available total revenue and sales data.

Results of modeling efforts show a total conservation in the range of 16% to 27% across the four cases with 10% to 24% attributed to price and income effects. The governor's call for voluntary emergency conservation during the 1976-77 gas emergency had little if any effect on consumers. Furthermore, the modeling results provide evidence that President Carter's tax rebate plan has had little if any effect.

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1. INTRODUCTION

The early 1970s marked the end of a long period of growth in residential space-heat demand in the U.S. This growth primarily had been due to increasing real income and decreasing real price of fuels and power. For example, Strout [1] examined space-heat demand for the period from 1935 through 1959 using national-level data and found in that period that disposable real per capita income increased 85% and a deflated price index for space heat decreased 27%. At the same time, per U.S. resident space heat increased 63%. The October 1973 Arab oil embargo signaled the end of such growth and the beginning of energy conservation.

While the oil embargo had little immediate effect on natural gas consumption in Ohio, by 1974 natural gas supply conditions were deteriorating markedly. In response, moratoriums on new customer hookups were declared, some nonresidential natural gas customers experienced peak-day and volumetric supply curtailments, and at the national level, natural gas prices were permitted to rise sharply. Thus, a general awareness of the conservation issue came into being.

Then, much of the eastern U.S. experienced the coldest-on-record 1976-77 heating season. This precipitated the 1976-77 gas emergency in which Ohio was among the hardest hit states. Portions of the industrial and commercial sectors were curtailed to plant protection levels resulting in significant unemployment impacts (see von Rabenau and Gorr [2]). On January 17, 1977, the governor of Ohio declared an energy emergency and called for voluntary emergency conservation measures in the residential and commercial sectors. Subsequently, schools were closed in Columbus, Ohio, to save additional natural gas. Awareness of the conservation issue was, perhaps, at its peak during this period.

The following 1977-78 heating season also was very cold, but supply conditions had improved and apparently conservation had contributed to a favorable supply and demand balance in Ohio. Perhaps contributing to the conservation experienced was the tax credit program of the Carter administration's National Energy Plan (see Hirst and Jackson [3]). This program included a 25% tax credit for retrofit conservation measures up to \$800 plus 15% credit on the next \$1,400 in the residential sector in effect from 1977 until 1984. While the tax credit program may have had some effect in early 1977, its effect is not likely to have been felt until the 1977-78 heating season due to delays in the program's approval. After a month or two into the 1977-78 heating season, the distribution utilities of Ohio, with the support of the Ohio Department of Energy, moved greatly to reduce or to eliminate volumetric curtailments to nonresidential customers. Since the 1977-78 heating season, there have been no volumetric curtailments in Ohio. Also, the moratoriums on new customer hookups were lifted, and new customers have been joining the system.

Thus, the 1970s produced pressures for conservation of residential natural gas in Ohio including economic incentives through natural gas price increases, a directive from the governor of Ohio to conserve on an emergency basis, a government incentive program through tax credit to implement conservation measures, and a great deal of information on the natural gas shortage and its impacts that appealed for voluntary conservation. The first of two purposes of this paper is to determine the extent of the resulting conservation and to evaluate the contribution of each conservation pressure or incentive on the total conservation effect. For this purpose, case studies were conducted on the four largest distribution utilities in Ohio. Monthly data on each utility have been collected for the period from January 1970 through March 1980. This provides a comprehensive study period encompassing before, during, and after implementation of conservation measures.

The second purpose of this paper is to introduce time-varying-parameter (TVP) models and the relatively new adaptive estimation procedure (AEP), due to Carbone and Longini [4,5], to the energy-modeling area. The TVP model has two related applications in evaluating energy conservation

measures, and for evaluation research in general. First, this paper applies the TVP model as a new technique for Exploratory Data Analysis, an approach to modeling promoted by Tukey [6]. In particular, AEP is used in a modeling process called "expansion modeling" that provides a staged, inductive approach to model building. Second, several researchers in the energy-modeling field have noted that the behavior of energy consumers is dynamic, resulting in constantly changing relationships in demand models (see for example, Chapman et al. [7], Verleger [8], Mayer and Horowitz [9], Bopp and Lady [10], and Sonnino [11]). Unfortunately, until recently, there have been no data analysis techniques directly able to handle the estimation of dynamic systems. For example, the ordinary least squares (OLS) regression technique, often used to estimate demand models, requires the restrictive assumption of unchanging (constant parameter) relationships between variables of a model. In contrast, AEP was designed to estimate changing relationships through the TVP model specification. This allows the construction of models not possible before.

Section 2 of this paper reviews conventional energy demand models, and section 3 introduces expansion modeling and AEP. Section 4 presents the four case studies of residential natural gas conservation in Ohio. Section 5 is a summary of the paper.

2. REVIEW OF ENERGY DEMAND MODELS

Strout [1], Chapman et al. [7], Verleger [8], Mayer and Horowitz [9], Bopp and Lady [10], Nelson [12], Saha and Stephenson [13], and others have used constant-parameter energy demand models identical or similar to the following:

General Energy Demand Model

$$Q_1/N = \beta_0 + \beta_1 P_1 + \dots + \beta_m P_m + \beta_{m+1} I/N + \beta_{m+1} D + \varepsilon \quad (1)$$

where m = number of substitutable fuels or power sources available for the case at hand;

Q_1 = aggregate demand per unit period for the particular fuel or power source under study, denoted as number 1 for notational convenience;

N = total number of customers generating Q_1 ;

P_j = unit price of fuel or power source j ;

I = income per unit period of the N customers being served;

D = sum of heating or cooling degree days per unit period;

β_j = parameter to be estimated; and error term.

Generally, the parameters of model (1) are estimated using the OLS method.

Model (1) is stated in linear form. Often this model also is used in log-linear form that has the appearance of model (1) but requires transformation by taking natural logarithms of all input data. Strout [1] states that the linear model is best suited for data of a short unit time period, say a month or less, while the log-linear model fits annual data best. Engineering models of heat requirements for buildings, based on

thermodynamic principles, yield a linear relationship between space heat and heating degree days (see Kusuda [14]). Section 4 uses a model retaining features of both the linear and log-linear forms.

The use of per customer demand, Q_1/N , in model (1) facilitates comparisons of model results across different geographic regions and time periods. As the price of fuel or power source 1 increases, it is expected that Q_1/N will decrease so that β_1 should be negative. In contrast, if a price of a substitute fuel or power source increases, it is expected that Q_1/N will increase so that β_2 through β_m should be positive. There is some debate as to whether marginal or average prices should be used in energy demand models, although marginal prices are often preferred on theoretical grounds. Many studies use average prices, however, since marginal price data are difficult to obtain, but average prices are calculated easily from total sales and revenue data. Marginal prices are estimated in section 4.2 using a new TVP model from total sales and revenue data. Section 4.2 provides a breakthrough in making marginal prices available from commonplace data.

Model (1), as stated, is considered to be a long-run model, since the prices of substitute fuels or power sources are included. The assumption is that the stock of appliances requiring fuel or power can change in the long run, and the prices of competing, substitute fuels or power sources will indicate the impacts on demand for fuel 1. Houthakker and Taylor [15] and Balestra [16] provide so-called "dynamic" models for the long run. At first, they include the stock of appliances as a variable in model (1) but then eliminate it by assumptions and algebraic manipulations. Their approach does not appear to add much to analysis. If only price P_1 , the own-price, is included in model (1), then the model is considered to be a short-run model with the stock of appliances fixed. Since substitutes have not been competitive overall to natural gas, it is reasonable to assume that the short run for residential natural gas consumption is a long period of time. Thus, prices P_2 through P_m are excluded from model (1) in the case studies of section 4.

Per capita disposable income is an indicator of consumption Q_1/N . As income increases, it is expected that consumption will increase so that $m+1$ should be positive in model (1). Heating degree days for a given day are calculated as

$$D = \text{maximum} \{0, 65-T\}$$

where T = daily average or midrange temperature ($^{\circ}\text{F}$).

The value "65" used in the calculation of D is actually a behavioral coefficient that should be estimated from historical data, since weatherization and lowered thermostat settings reduce this coefficient. Mayer and Horowitz [9] have attempted to estimate this coefficient over time but encountered severe limitations. Preliminary investigations on the Ohio case studies show that variations in this coefficient away from the value "65" can be ignored safely.

The results of demand models, such as model (1), are often reported in terms of the dimensionless quantities, elasticities of demand. For model (1), an elasticity is the percentage change in fuel 1 consumption per customer caused by a 1% increase in a specified price, income, or other variable. Elasticities for the log-linear form of model (1) are simply the estimated values $\beta_1, \dots, \beta_{m+2}$. For the linear form, the elasticities are

$$\eta_{x_j} = \hat{\beta}_j x_j / (Q_1/N)$$

where η = elasticity of demand; and (2)

$$x_j = \text{jth variable of model (1).}$$

3. EXPANSION MODELING OF ENERGY DEMAND USING TIME-VARYING PARAMETERS

A straightforward estimation of model (1) for the case studies of this paper fails to produce useful results because the variables natural gas price and personal income are highly collinear in both original and logged forms. In particular, both of these variables increased during the study period so that their effects are not separable. As a result, signs of the estimated coefficients for these variables are not as expected in general, and usually one of the two variables is insignificant statistically. The purpose of an evaluation study such as this one, however, is to provide models with meaningful estimates of coefficients. Thus, it became necessary to attempt to go beyond conventional energy demand models and to search for new formulations to provide useful assessments of energy conservation pressures. This section, therefore, introduces an inductive modeling approach, one that builds on patterns discovered in the case study data, for this purpose. First, TVP models are introduced. These models provide the basis for a staged model-building procedure starting with aggregate models and then proceeding with detailing appropriate portions of the model as required. The second part of this section provides a top-down modeling process due to Casetti [17, 18] known as "expansion modeling." The use of TVP models as the basis for the top-level model is new.

3.1 Time-Varying-Parameter Models

In general terms, a TVP model is denoted

$$Y(t) = \beta_0(t) + \beta_1(t)x_1(t) + \cdots + \beta_m(t)x_m(t) + \varepsilon(t) \quad (3)$$

where $Y(t)$ = dependent variable at time t ;
 $x_j(t)$ = j th independent or explanatory variable;
 $\beta_j(t)$ = TVP of the j th independent variable; and
 $\epsilon(t)$ = unspecified model error term.

Energy demand model (1) is easily extended to this form by adding the t subscript to all model coefficients. The estimated model is denoted

$$\hat{Y}(t) = \hat{\beta}_0(t) + \hat{\beta}_1(t) x_1(t) + \cdots + \hat{\beta}_m(t) x_m(t)$$

Two self-adaptive data analysis techniques are available to provide the $\hat{\beta}_j(t)$ estimates: AEP due to Carbone and Longini [4,5], and Least Mean Squares due to Widrow et al. [19] and further developed by Wheelwright and Makridakis [20]. Both techniques are based on feedback principles and were developed heuristically to yield good performance characteristics. Bretschneider and Gorr [21] have shown some theoretical properties of these and related techniques. This paper uses AEP.

The approach of AEP is as follows. At time t , a one-step-ahead forecast is made using known parameter estimates from $t-1$ and independent variable data from t :

$$Y^F(t) = \hat{\beta}_0(t-1) + \hat{\beta}_1(t-1) x_1(t) + \cdots + \hat{\beta}_m(t-1) x_m(t)$$

Next, the forecast error $Y(t) - Y^F(t)$ is calculated. Through the direction of a simple pattern recognizer in the AEP filter, small adjustments are made to $\hat{\beta}_j(t-1)$ to yield $\hat{\beta}_j(t)$ ($j = 0, \dots, m$). The new coefficients reduce, but do not eliminate, the forecast error $Y(t) - Y^F(t)$. This process is repeated recursively with each new observation so that estimated parameter paths drift over time to capture dynamic effects of the system under study.

3.2 Expansion Modeling

Work initiated by Carbone and Gorr [22] suggests that TVP models are useful in inductive modeling procedures. Inductive modeling has been

legitimized in recent years by the Exploratory Data Analysis movement introduced by Tukey [6]. The idea here is that analysts should look for patterns in available data as an aid to modeling a phenomenon under study. This is in sharp contrast to the classical modeling approach that requires that analysts determine the model completely on theoretical grounds and then obtain data needed for estimation of the model.

The modeling process introduced and applied in this paper involves a series of modeling efforts in which the goal is to introduce a sufficient number of appropriate variables to eliminate systematic variation in TVP paths. This results in a final, constant-parameter model. Three steps are required: (1) a top-level, aggregate TVP model is specified and estimated; (2) each TVP with systematic variation is expanded (or modeled) through the introduction of new independent variables, and (3) all parameters of the expanded model are estimated simultaneously. Each of these steps is discussed in some detail below.

An example of a top-level model is the commonly used heating degree day model supplemented with the TVP specification:

$$Q_1(t)/N(t) = \alpha_0(t) + \alpha_1(t)D(t) + \varepsilon(t) \quad (4)$$

where $\alpha_0(t)$ and $\alpha_1(t)$ are top-level TVPs.

As is true for all top-level models, the terms of model (4) provide mutually exclusive and exhaustive categories for sources of variation in the dependent variable. These are the non-space-heating, $\alpha_0(t)$, and the space-heating, $\alpha_1(t)D(t)$, terms. Another property of top-level models is that some of their parameters are expected to vary over time due to identifiable independent variables not yet included in the model. These are the variables used to expand the TVPs in the second step. In model (4), systematic declines in $\alpha_0(t)$ and $\alpha_1(t)$ are expected in the Ohio case studies due to conservation pressures and incentives.

The results of top-level models are useful in directing subsequent modeling steps. First, the direction, magnitude, and timing of parameter variation are informative. Such patterns are suggestive of which additional variables might be helpful in expanding $\hat{\alpha}_j(t)$ s in the second step of this procedure. If a particular $\hat{\alpha}_j(t)$ does not vary significantly over time, then it is not necessary to collect data on additional variables to expand that $\hat{\alpha}_j(t)$. This is an economy that might not otherwise be had. Second, if time and resources are short, the results of the top-level model are inexpensively and quickly obtainable and may be sufficient when coupled with judgment to make necessary decisions.

After the top-level model has been estimated and studied, the next step is to expand the set of $\hat{\alpha}_j(t)$ s that has significant, systematic variation as follows:

$$\hat{\alpha}_j(t) = \hat{\gamma}_{0j} + \hat{\gamma}_{1j}x_{1j}(t) + \dots + \hat{\gamma}_{mj}x_{mj}(t) \quad (5)$$

where m = number of independent variables in the j th expansion,

x_{ij} = i th independent variable of the j th expansion; and

$\hat{\gamma}_{ij}$ = parameter i of the j th expansion.

While the expansion model (5) is stated as a constant-parameter, general linear model, any functional form as needed may be used and any estimation technique may be employed to estimate coefficients $\hat{\gamma}_{ij}$. Of primary importance here is the task of determining which variables x_{ij} are useful in explaining $\hat{\alpha}_j(t)$ and which functional form is most appropriate (linear, multiplicative, etc.). The estimates $\hat{\gamma}_{ij}$ generally are not themselves used in the final model, obtained in step 3.

The expansion step carries out the analytical process started with the top-level model. The top-level model provides diagnostic information on which $\hat{\alpha}_j(t)$ s to expand and what independent variables might be useful. The expansion step then allows the analyst to focus on one component of the total system at a time, in terms of firming up the form of the final model. In step 3, the model form as in (4) is expanded by substituting for $\hat{\alpha}_j(t)$ the model forms found successful in model (5).

4. CASE STUDIES

This section presents case studies on natural gas conservation in the residential sectors of the four largest distribution utilities in Ohio: Cincinnati Gas and Electric, Columbia Gas of Ohio, Dayton Power and Light, and the East Ohio Gas Company. Appendix A contains the raw data for these cases including 123 monthly observations for each utility covering the period from January 1970 through March 1980. Also included in the appendix are discussions of various aspects of the data.

Table 1 contains descriptive statistics on the data of appendix A. It is seen in this table that mean per customer consumption is about the same for three of the utilities, but the typical customer of East Ohio Gas consumes about 20% more natural gas than those of the other companies. The heating degree day data of table 1 provide a partial explanation of this difference. The heating season of East Ohio Gas customers, in northern Ohio, is about 10% colder than those of the other utilities, and since space heat is the dominant portion of residential natural gas consumption, this accounts for nearly 10% of the 20% difference. There is no explanation available at this time for the remaining 10% difference. The average price variable in table 1 was obtained by dividing monthly revenue by monthly consumption. The range of prices shows the price increases over time, from about \$0.85 to \$3.60 per mcf--an increase of about 425% in 10 years. At the same time, national personal income rose from $\$779 \times 10^9$ to $\$2,070 \times 10^9$, only an increase of 265% as compared to the increase in natural gas prices. Note that national personal income data are used because state and city-level income data were not available on a monthly basis. On an annual basis, the national and Ohio data have a correlation of 0.99, so that the national-level variable provides useful information.

TABLE 1

DESCRIPTIVE STATISTICS BY DISTRIBUTION UTILITY FOR MONTHLY DATA
CONCERNING RESIDENTIAL NATURAL GAS CONSUMPTION

Variable	Company*	Units	Sample Size	Mean	Standard Deviation	Minimum Value	Maximum Value
Natural Gas Consumption per Customer	CG&E	mmcf	123	0.0130	0.00939	0.00265	0.0327
	CGO		123	0.0142	0.00953	0.00322	0.0339
	DP&L		123	0.0131	0.00930	0.00263	0.0345
	EOG		122	0.0168	0.01089	0.00395	0.0388
Heating Degree Days	CG&E	Total	123	457	409.5	0.00	1399
	CGO	Monthly	123	491	425.5	0.00	1462
	DP&L	°F	123	488	426.7	0.00	1384
	EOG		122	531	434.4	3.00	1398
Natural Gas Average Price	CG&E	\$/mcf	123	1.84	0.86	0.87	3.58
	CGO		123	1.91	0.84	0.86	3.74
	DP&L		123	1.81	0.81	0.75	3.49
	EOG		122	1.70	0.70	0.79	3.64
National Personal Income	--	\$10 ⁹	123	1280	377	779	2070
Consumer Price Index	--	1967 = 100	123	159	34.7	113	240

*CG&E = Cincinnati Gas and Electric
 CGO = Columbia Gas of Ohio
 DP&L = Dayton Power and Light
 EOG = East Ohio Gas

Sources: Consumption, revenue, and number of customers data were obtained directly from the utilities; degree day data were obtained from Climatological Data, National Climatic Center, Asheville, North Carolina; income data were obtained from Survey of Current Business, U.S. Department of Commerce; and Consumer Price Index data were obtained from CPI Detailed Report, U.S. Department of Labor

4.1 Top-Level TVP Models

This section provides estimates of top-level model (3). In keeping with the philosophy of Exploratory Data Analysis, several graphs are presented first that guide model construction. The Columbia Gas case is used to illustrate supporting concepts, but top-level modeling results are given for all cases.

Figure 1 depicts the time series of monthly per customer consumption for Columbia Gas, and figure 2 shows the corresponding time series of degree days. Clearly, these figures show very similar patterns due to the strong seasonality of degree days. Heating degree days are zero or nearly zero in July, August, and September so that consumption from these months represents non-space-heat consumption. Note that consumption peaks in January and February are 10 times higher than the summer troughs.

Further examination of figures 1 and 2 shows some indication of conservation. First, note that the non-space-heat troughs of figure 1 decline noticeably starting in 1976 or 1977. Also, figure 2 has as its highest peak degree days from the 1976-77 heating season, the coldest winter on record and significantly colder than the winters of 1969-70 through 1971-72. Note, however, that the 1976-77 consumption peak in figure 1 is about the same as those for the 1969-70 through 1971-72 heating seasons. This is a clear indication of conservation in the space-heat component of consumption.

The scatter plot of figure 3 more clearly depicts the relationship over time between per customer consumption, degree days, and conservation for Columbia Gas. Here, the plot character is the last digit of the year of observation with X's drawn for 1980 to distinguish 1980 from 1970. At any point in time, figure 3 shows that approximately a linear relationship exists between per customer consumption and degree days. As time passes, the slope and intercept of this relationship decrease continuously; thus, the TVP model (4) is appropriate.

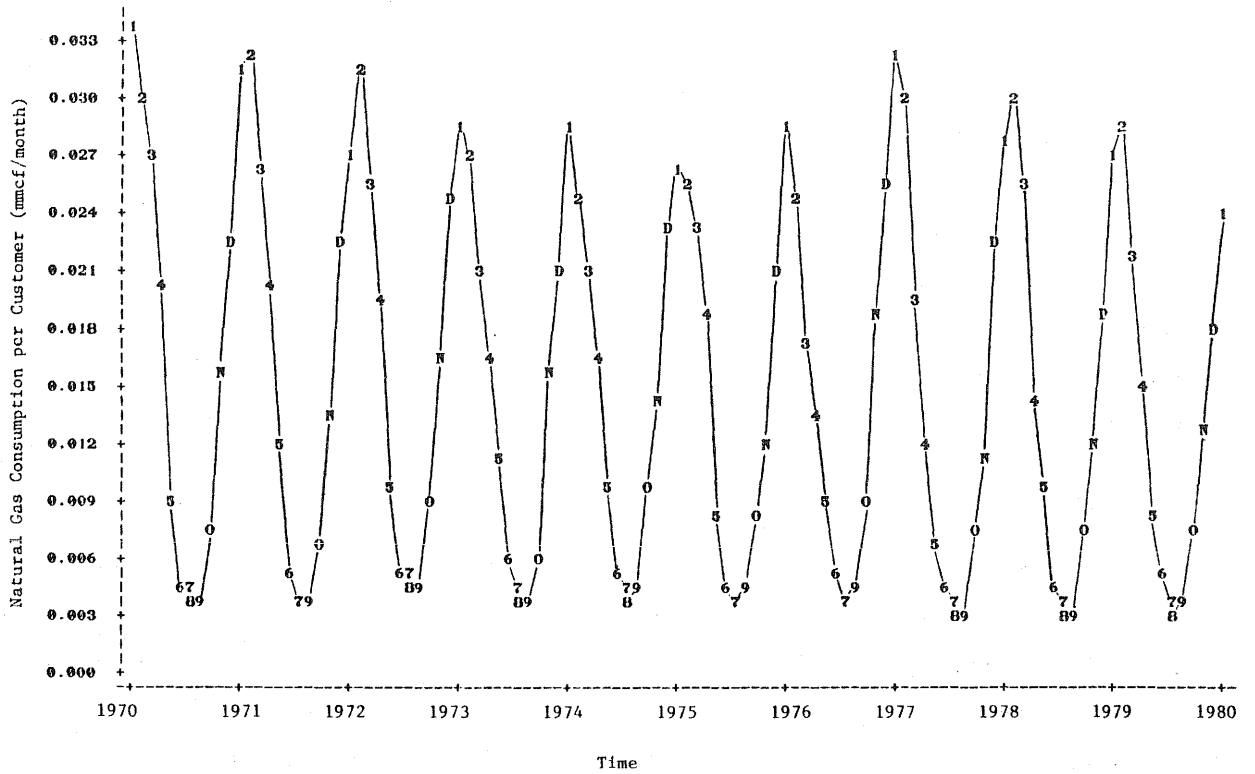


Fig. 1 Time series plot of residential natural gas consumption per customer: Columbia Gas of Ohio

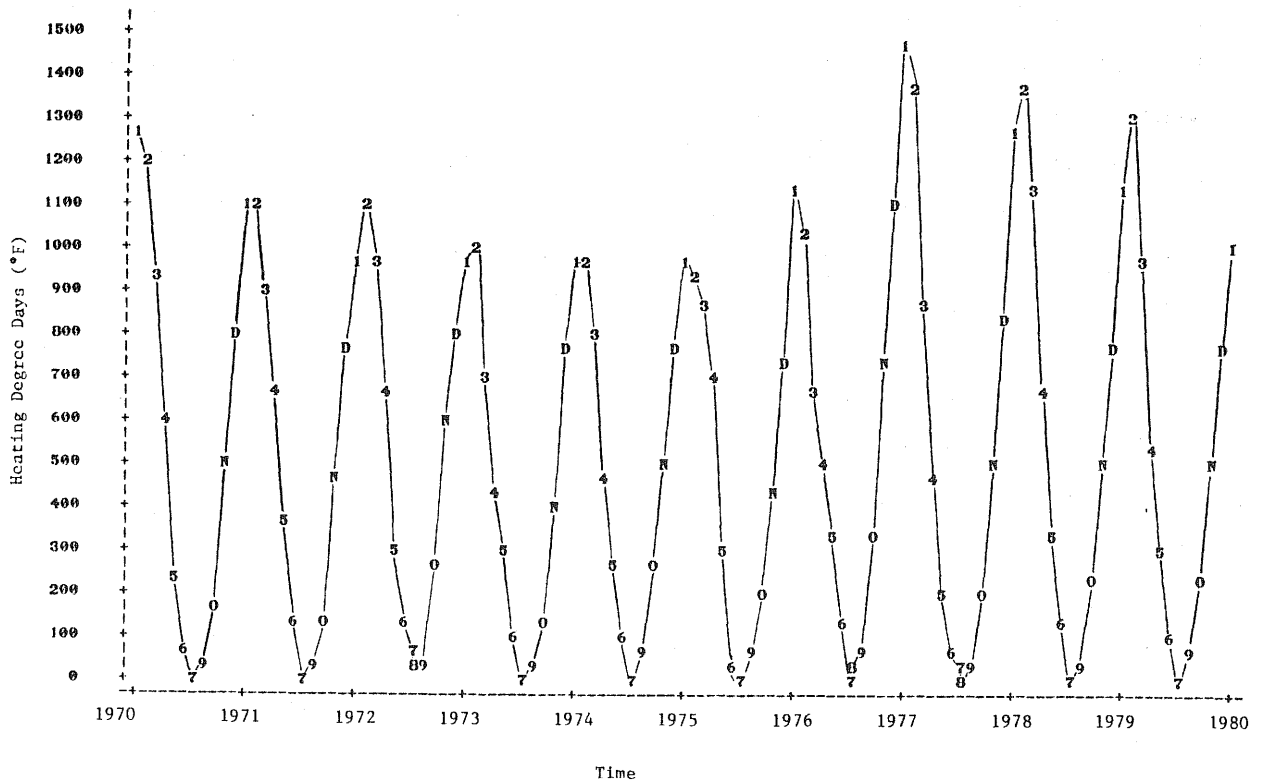


Fig. 2 Time series plot of heating degree days from the Port Columbus International Airport

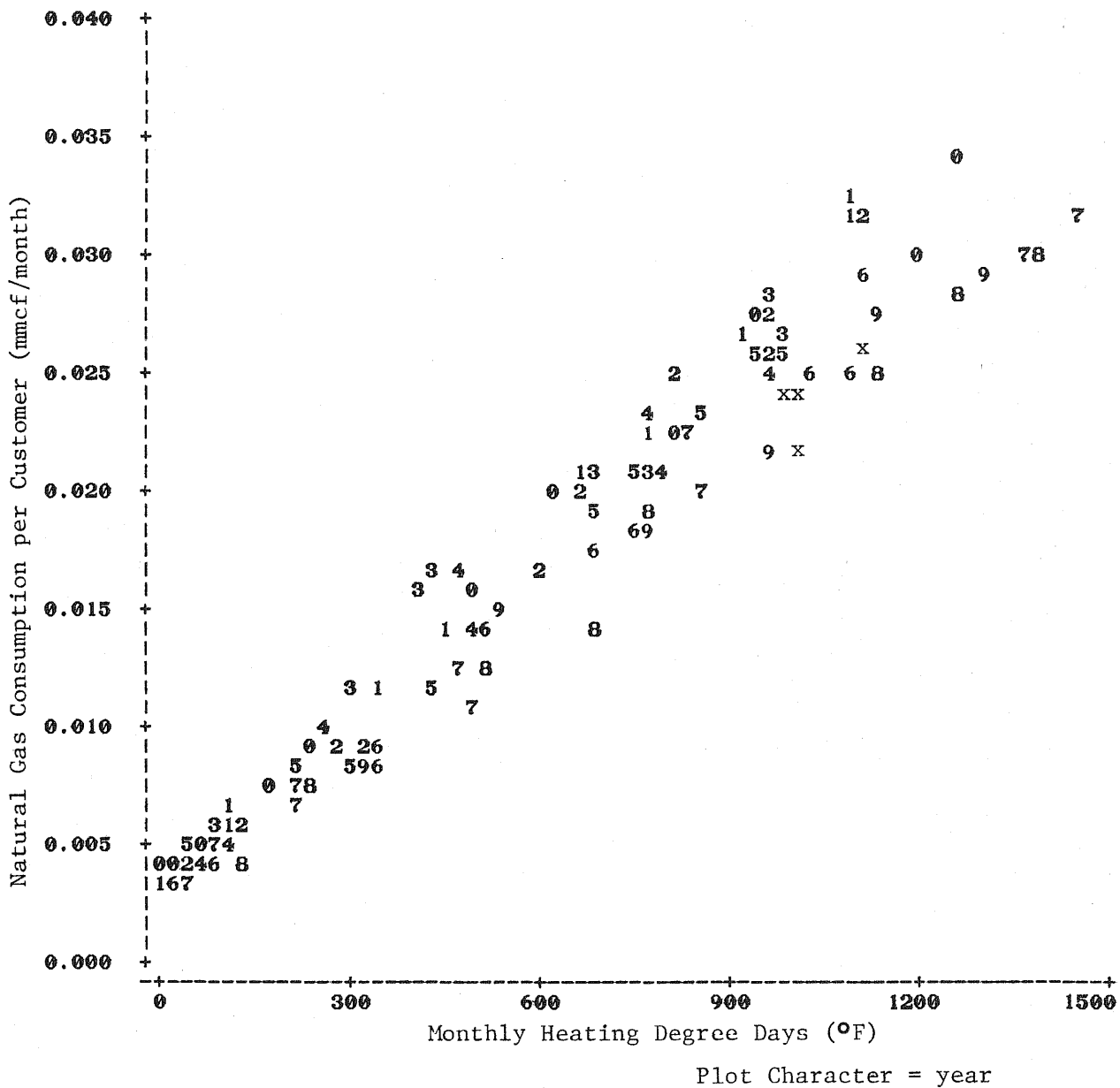


Fig. 3 Scatter plot of residential natural gas consumption per customer versus heating degree days: Columbia Gas of Ohio

Values selected for various constants and initial values for use in AEP calculations included damping factor = 0.2, smoothing constant for means of explanatory variables = 0.01; correction limit = 0.2; $\alpha_0(0) = \alpha_1 = 1$; and the number of forward and backward cycles through the data = 40. The damping factor and correction limit are roughly twice as large as values generally found to be satisfactory and were necessary to remove all systematic patterns from residuals of model (4). As a result, the estimated parameter paths for model (4) presented in figures 4 through 11 are less smooth than usual. The standard error of estimate averages 0.0013 for these AEP models. Compared to the average per customer consumption of 0.014 mmcf/mo. for the four utilities, this standard error indicates an excellent fit of model (4) to the data.

Figures 4 and 5 present estimated parameter paths for the intercept $\hat{\alpha}_0(t)$ and slope $\hat{\alpha}_1(t)$ as estimated for Cincinnati Gas and Electric. The intercept declined steadily until about January 1977, after which it remained approximately constant. The total decline was about 16%. The slope increased erratically until September 1972, declined until about March 1973, remained fairly constant until April 1977, then declined sharply until April 1977 after which it remained roughly constant. The overall decline for the slope from January 1974 to January 1978 was 17%. Using a value of 900°F for monthly degree days and parameter estimates from January 1974 and January 1978, per customer consumption is estimated to have declined 16% in total. In summary, figures 4 and 5 clearly show the impacts of conservation measure for Cincinnati Gas and Electric. The sharpest impacts were produced during the severe 1976-77 heating season, but shortly thereafter conservation was arrested or reversed. Apparently, the tax credit program had little or no effect on conservation according to the TVP model, since the program was not approved until April 1977.

Next, figures 6 and 7 depict parameter paths for Columbia Gas. The intercept behaved erratically until January 1974, after which it declined rapidly until early 1977. Since then, it increased slightly. The decline

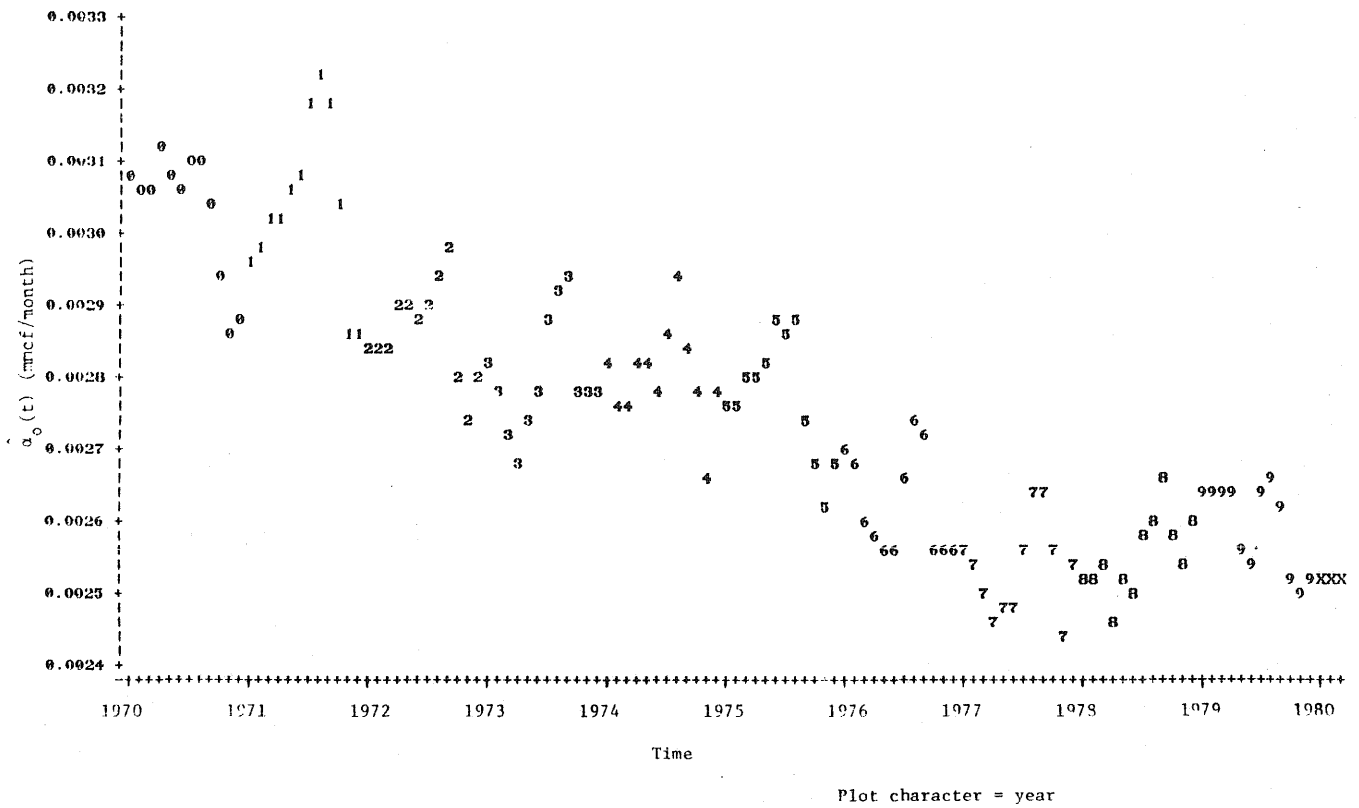


Fig. 4 Parameter path for the intercept term of the linear heating degree day model: Cincinnati Gas and Electric

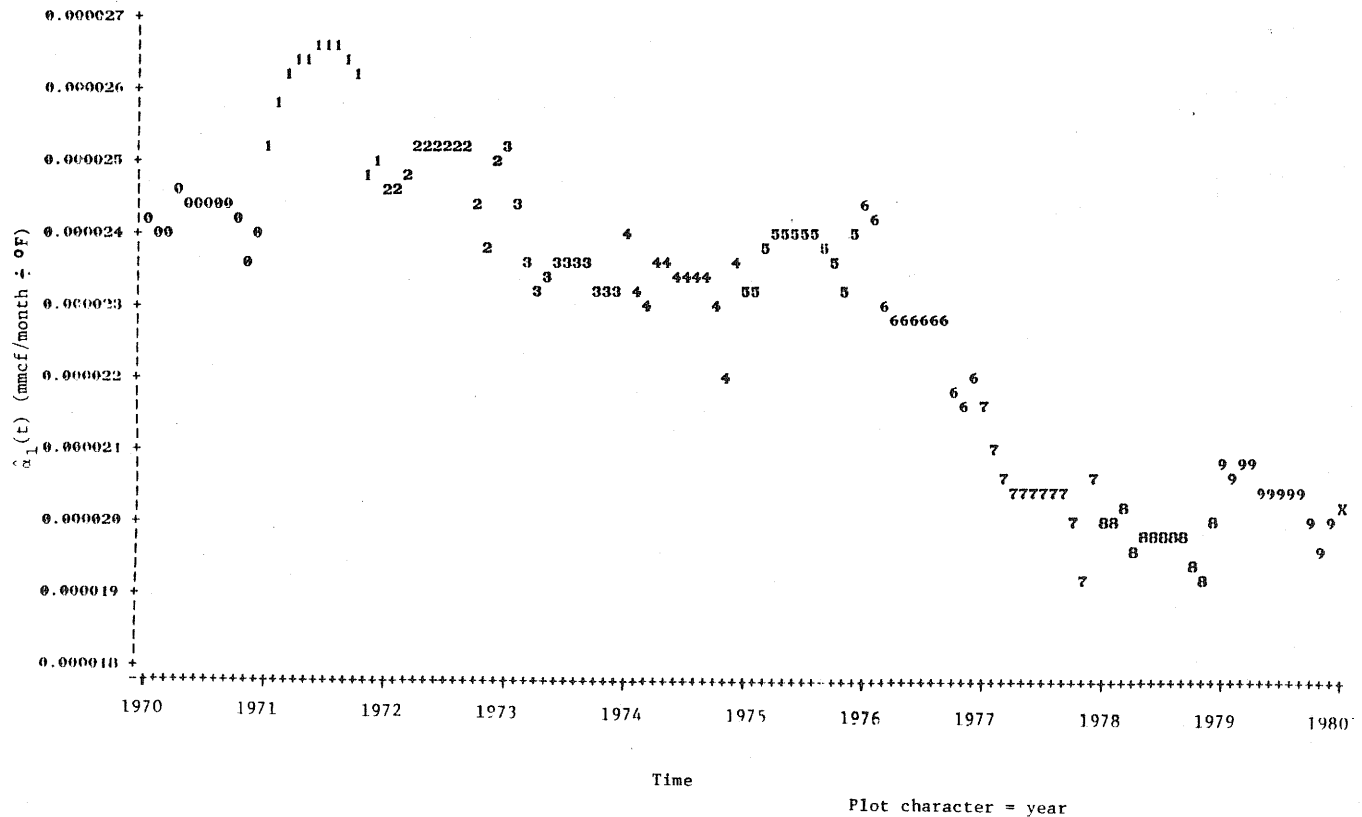


Fig. 5 Parameter path for the heating degree day term of the linear heating degree day model: Cincinnati Gas and Electric

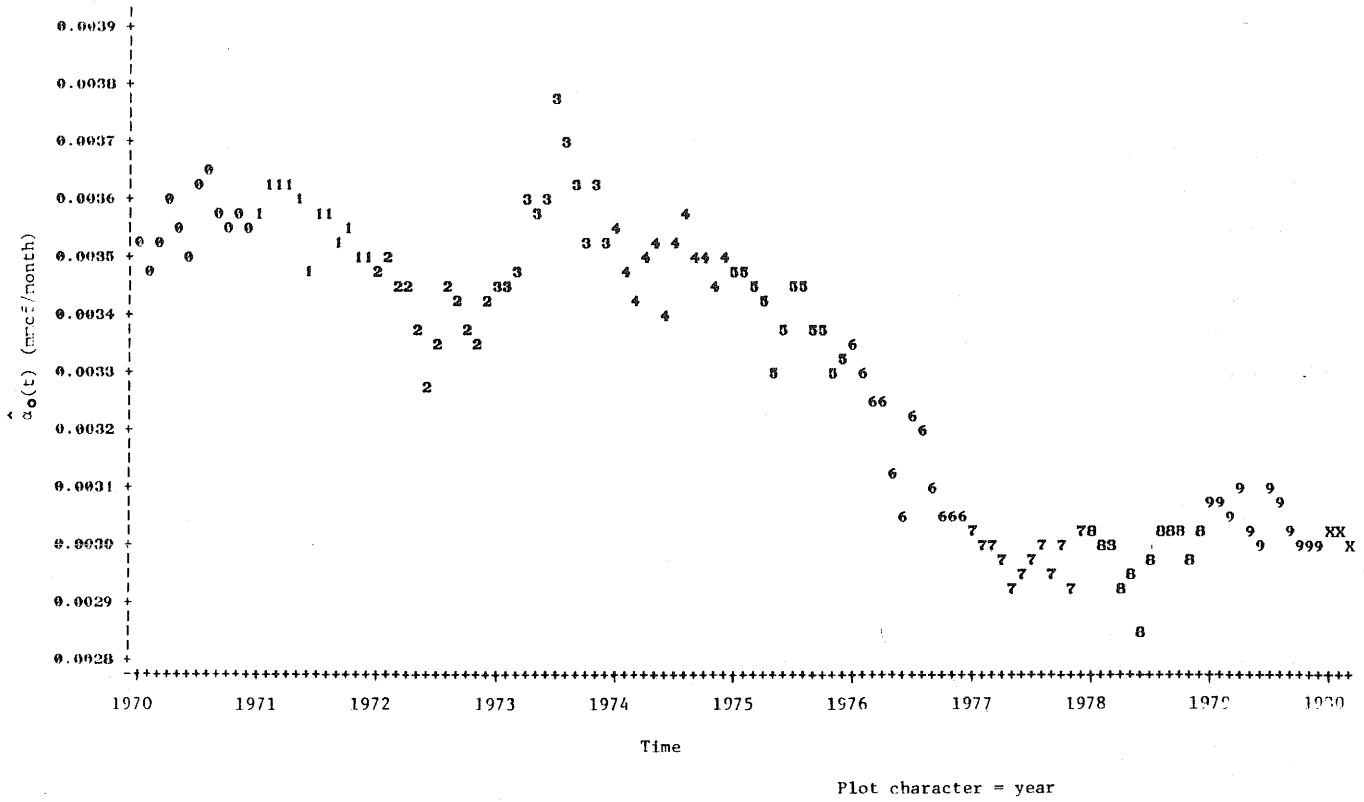


Fig. 6 Parameter path for the intercept term of the linear heating degree day model: Columbia Gas of Ohio

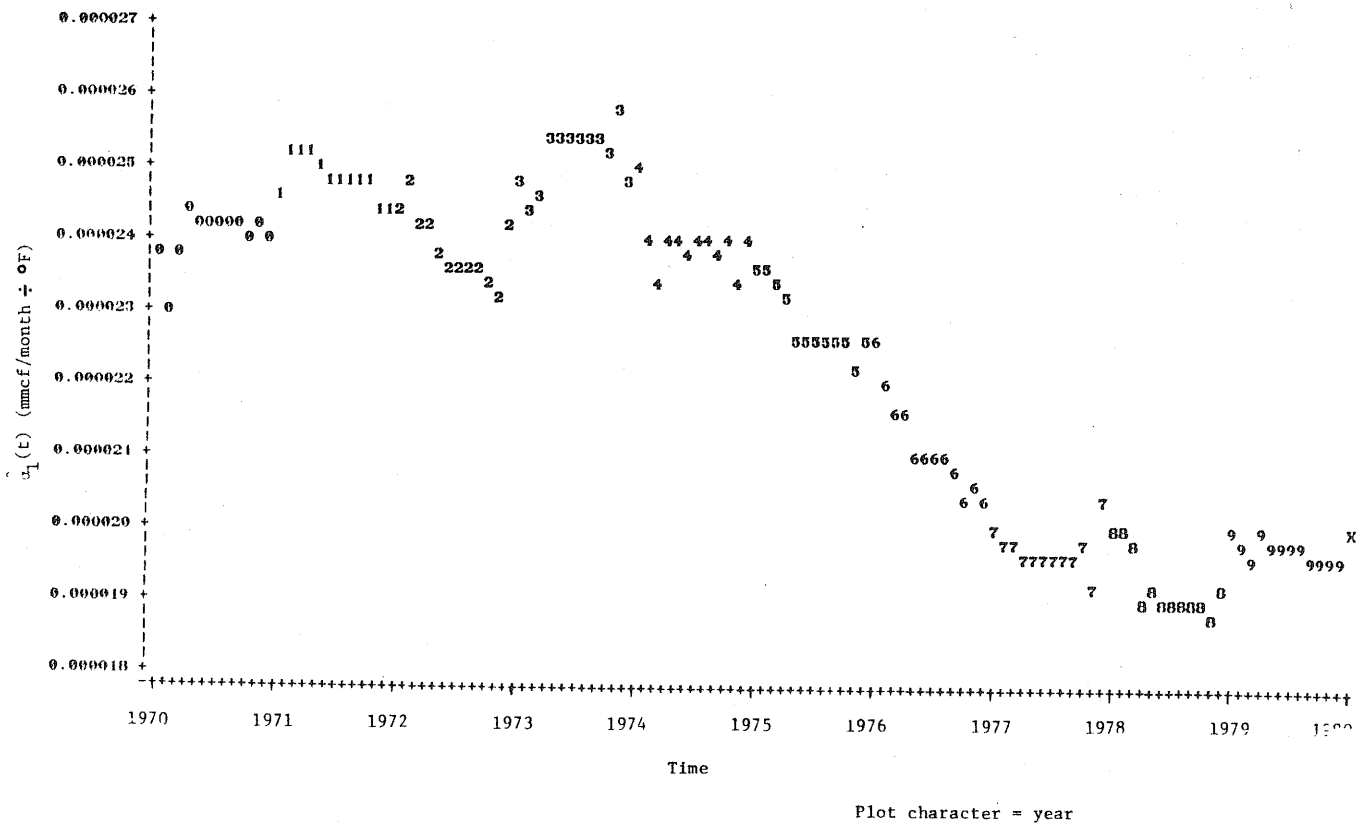


Fig. 7 Parameter path for the heating degree day term of the linear heating degree day model: Columbia Gas of Ohio

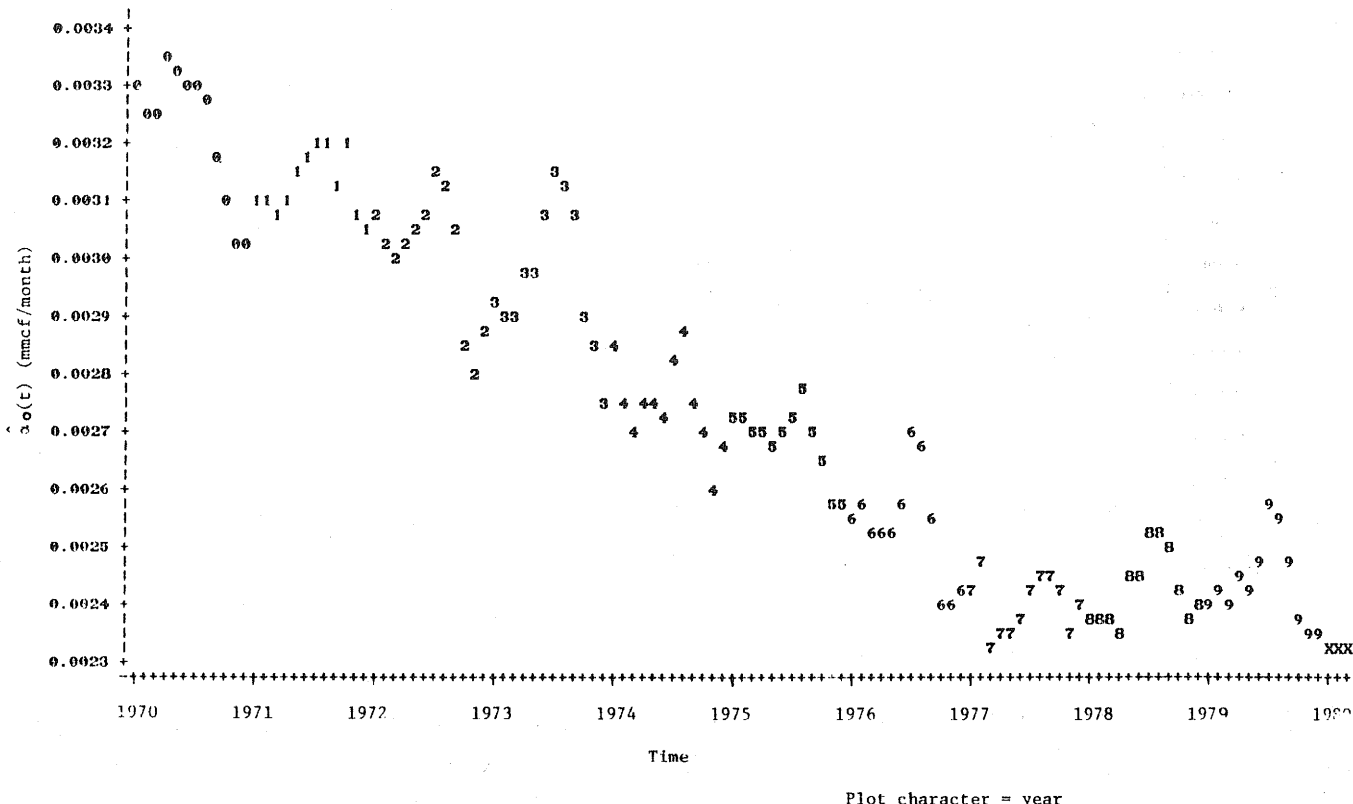


Fig. 8 Parameter path for the intercept term of the linear heating degree day model: Dayton Power and Light

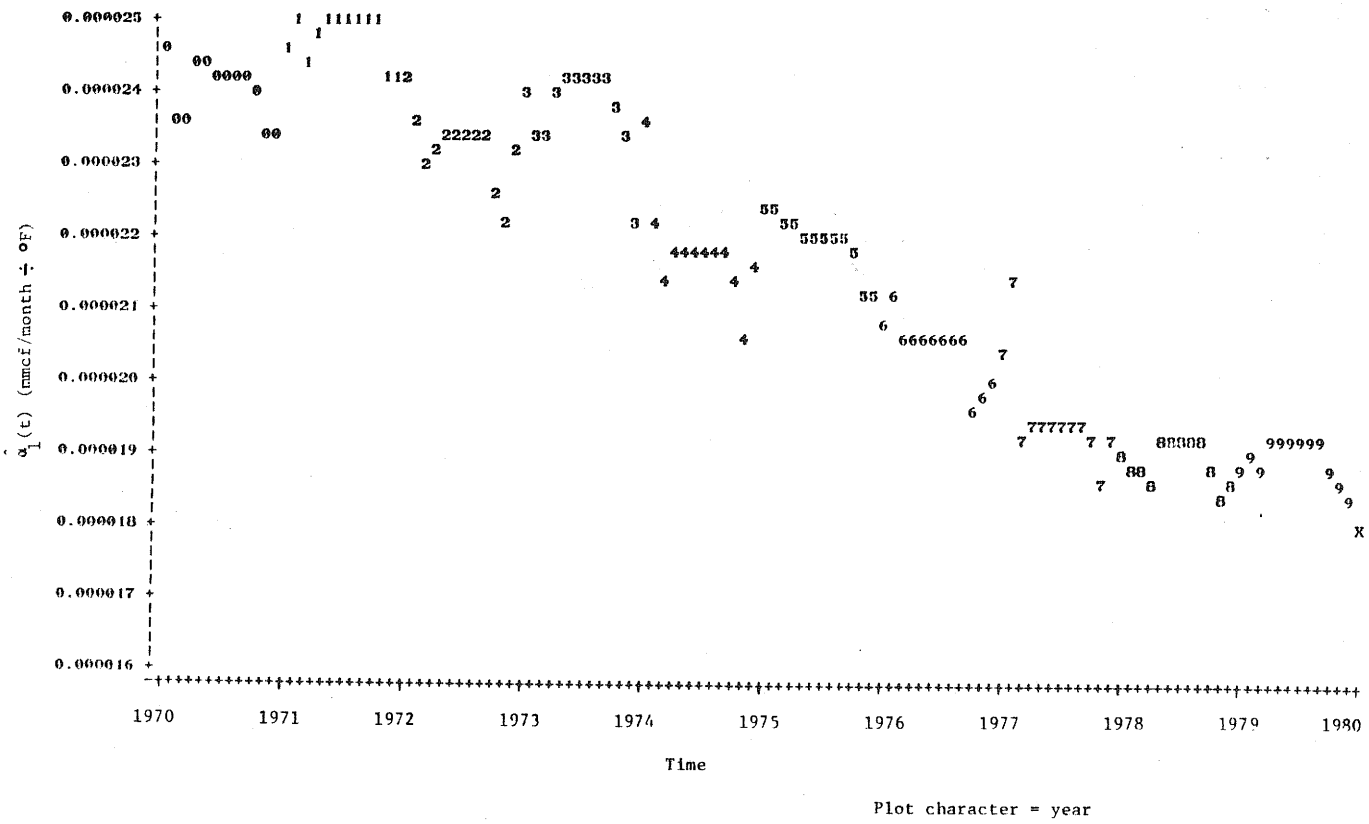


Fig. 9 Parameter path for the heating degree day term of the linear heating degree day model: Dayton Power and Light

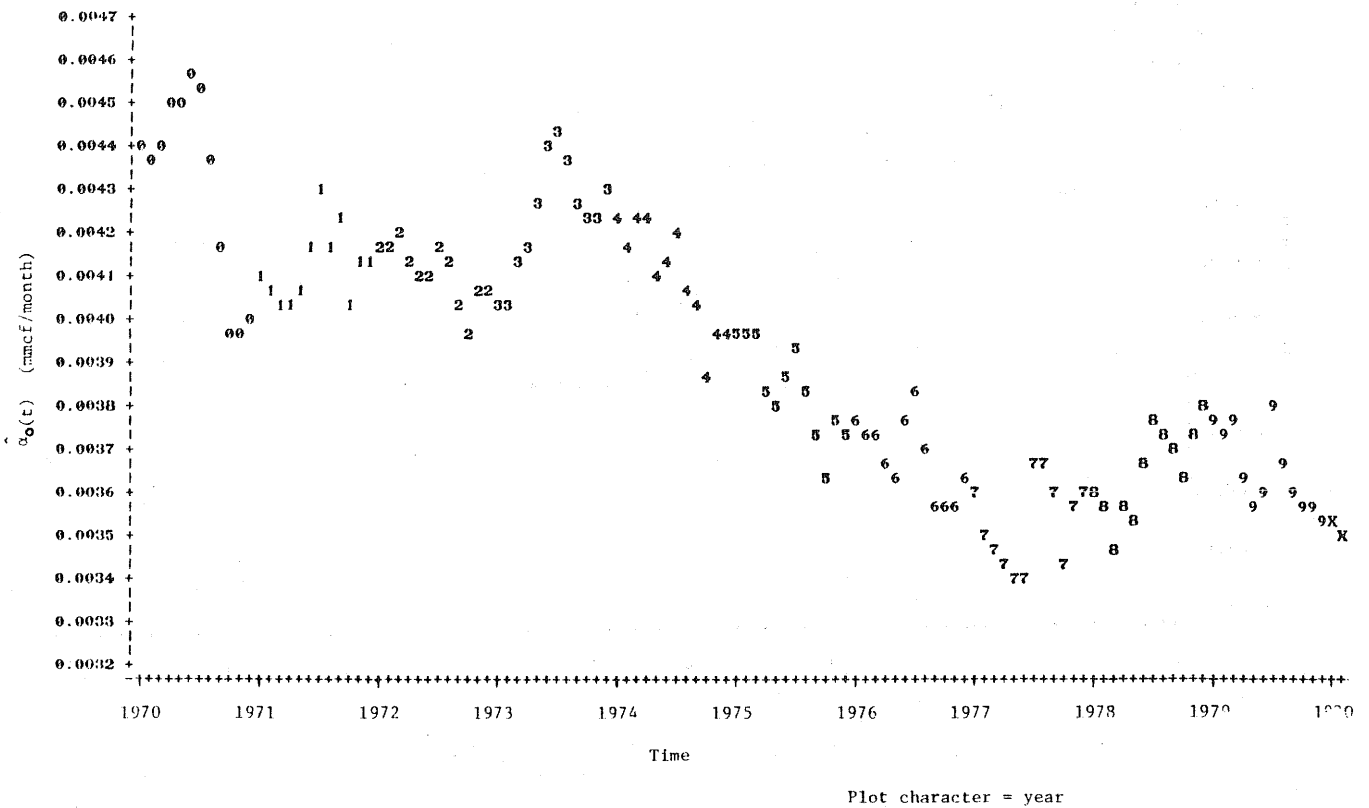


Fig. 10 Parameter path for the intercept term of the linear heating degree day model: East Ohio Gas

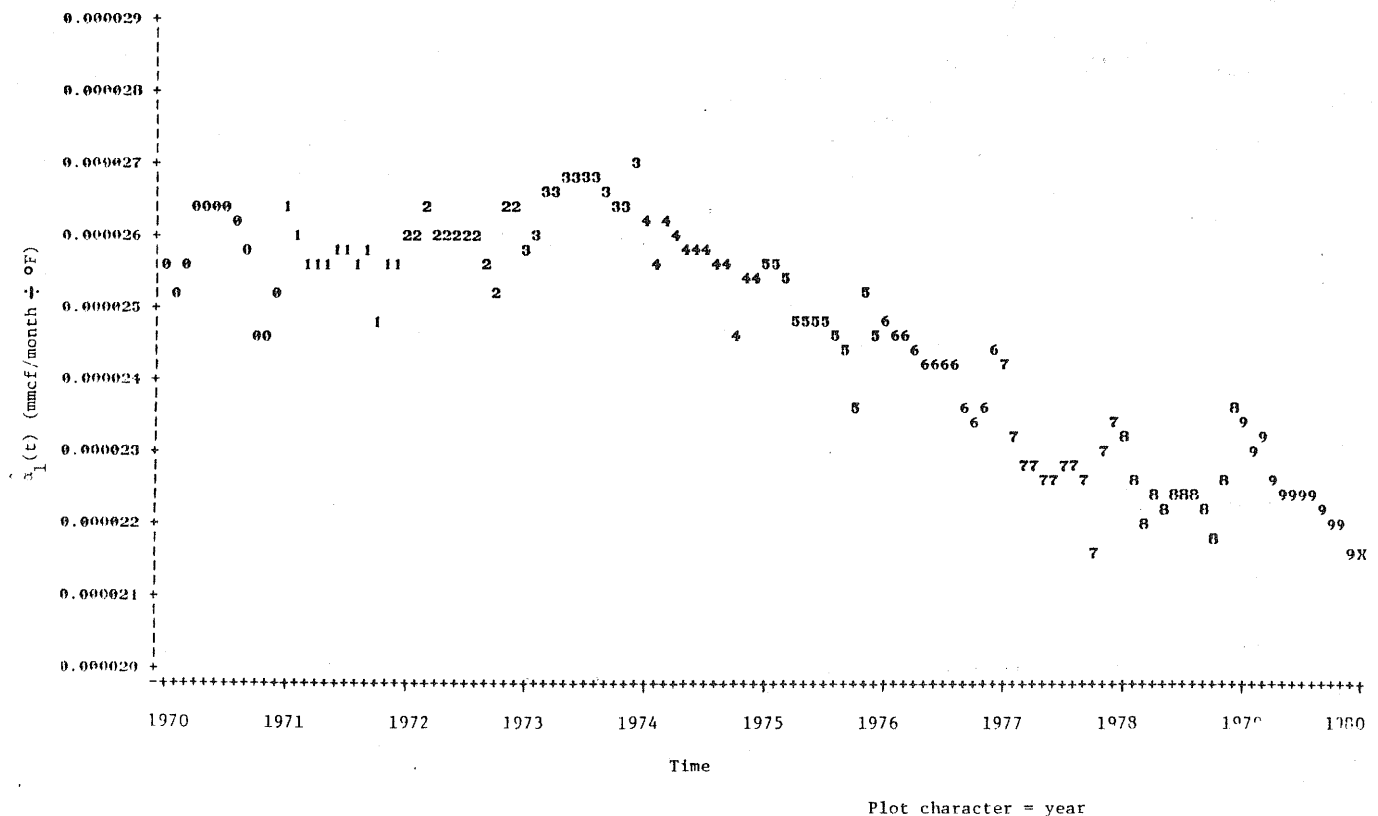


Fig. 11 Parameter path for the heating degree day term of the linear heating degree day model: East Ohio Gas

in the slope from January 1974 until January 1978 was 21%. Again using 900°F and January 1974 and 1978 parameter estimates, the total decline in consumption was 20%. These figures for Columbia Gas show patterns somewhat different from those of Cincinnati Gas and Electric; for example, the decline from 1974 to 1977 was more steady for Columbia Gas. Again, the tax credit program appears to have had little or no conservation effect.

Figures 8 and 9 portray the case of Dayton Power and Light. The patterns in these figures are quite similar to those of Cincinnati Gas and Electric; however, the declines in parameters are much stronger. This is probably due to the fact that Dayton Power and Light has the highest percentage of residential customers in the total mix of customers for the four case studies. As there were proportionally fewer nonresidential customers to curtail during supply shortages, this put increased pressure on residential customers to conserve. The intercept declined 28%, and the slope declined 22% over the period from January 1974 until January 1978. Consumption fell 23% using 900°F and January 1974 and 1978 as before. Once again, we have evidence that the tax credit program was ineffective.

Last, figures 10 and 11 give the case of East Ohio Gas Company. The patterns in these figures are very similar to those of Columbia Gas. The intercept term for East Ohio Gas declined 13%, while the slope declined 14% over the period from January 1974 to January 1978. Overall, the decline in per customer consumption was 14% using calculations as above. As before, the tax credit program appears ineffective.

In summary of this section, the parameter paths from estimates of top-level model (4) via AEP provide explicit models of structural changes in the non-space-heat and space-heat components of each case study. The two smaller companies in the southwest of Ohio, Cincinnati Gas and Electric and Dayton Power and Light, are quite similar in behavioral patterns, although the latter company shows higher levels of conservation. The two largest utilities, East Ohio and Columbia Gas, show similar patterns with steady improvements in conservation from 1974 to 1977. As

will be seen in the next section, natural gas prices and income were favorable for slight growth in consumption until 1974, after which pressures steadily increased for conservation until they stopped in 1977.

The tax credit program, if it had an effect on conservation, would have shown up as declines starting in the 1977-78 heating season in the degree day paths of figures 5, 7, 9, and 11. To the contrary, the paths leveled off or increased during this period for all four utilities. This provides strong evidence that the tax credit program was ineffective as a conservation measure.

It should be noted that model (4) should provide excellent short-run forecasts using the AEP forecasting procedures due to Bretschneider et al. [24], Bretschneider [25], and Bretschneider and Carbone [26]. Roblee [27] is pursuing the short-run forecasting issue including all three sectors (residential, commercial, and industrial) for the cases of the four distribution utilities of this paper.

4.2 The Expansion Modeling Step

Four independent variables are under consideration to expand model (4):

- P_1 = marginal price of natural gas;
- I = personal income;
- E = dummy variable for the 1976-77 natural gas emergency ($E = 1$ for January through April 1977 and 0 otherwise); and
- T = dummy variable for the tax credit program ($T = 1$ after April 1977 and 0 otherwise).

Since a straightforward estimation of model (1) showed that P_1 and I are not separable due to multicollinearity problems, it was decided to attempt various combinations of these two variables including the ratios P_1/I and I/P_1 . Before pursuing this subject further, it is necessary to address

the subject of average versus marginal prices of natural gas, which is done in section 4.2.1.

The dummy variable T was shown to be ineffective in impacting Q_1/N in the top-level models of section 4.1. Thus, this variable is not included in the expansion of model (4). Furthermore, dummy variable E affects only 2 months of the 120-month study period, and these months are at the end of the period of significant decline in the parameters being expanded. Thus, E is not included in the expansion step of the modeling process either.

One notable missing variable in this model is a measure of conservation pressures for voluntarism. Campaigns to inform consumers of the need to preserve scarce resources, impacts of fuel shortages in the commercial and industrial sectors, goals to become less dependent on foreign supplies of fuels, etc., all can have impacts on conservation. Unfortunately, it is difficult to obtain a measure of voluntarism pressures. Perhaps content analysis could be applied to provide such a measure, but this has not been attempted in this paper. Even if such a measure were available it probably would not be useful in the sense that it likely would be collinear with other explanatory variables. Peck and Doering [28] and Walker [29] provide some results on voluntarism in energy conservation.

4.2.1 TVP Model for Marginal Price

Natural gas costs to the consumer are made up of a fixed monthly charge plus a variable cost calculated as the product of the unit price times the volume of natural gas consumed. Declining block rates are not used for residential natural gas customers of any of the utilities examined in this paper, so that the unit price, constant at any point in time, is the marginal price. While marginal price data are desirable in modeling consumption, such data are not readily available. For example, Columbia Gas of Ohio negotiates separate price tariffs with over 700 municipalities in Ohio. In this case, it is impractical for an independent researcher to determine an aggregate residential sector marginal price using accounting

methods; however, it is possible to estimate such marginal prices using a TVP model as explained below in this section. First, it is instructive to demonstrate the inadequacy of average cost data.

Data available in appendix A allow the direct calculation of average price as follows:

$$AP_1(t) = R_1(t)/Q_1(t)$$

where $AP_1(t)$ = average price,

$R_1(t)$ = monthly revenue from residential-sector sales; and

$Q_1(t)$ = monthly residential-sector sales.

Figure 12 gives a time series plot of the ratio $I(t)/AP_1(t)$ for Columbia Gas. The long-range trend of this plot shows conditions conducive to growth in consumption until 1974, pressures for conservation from 1974 until 1977, and no change thereafter. The seasonality of this plot, reflecting high average price in the summer and low average price in the winter, provides false information on consumption impacts of the ratio $I(t)/AP_1(t)$. Natural gas consumption is low in summer so that most of the cost is due to the fixed cost component of total cost. Thus, pricing policies using fixed and variable components of cost cause the seasonality of figure 12. Model (1), however, attributes low consumption in summer to consumer behavior in response to high average natural gas costs in summer; This is incorrect. This results in average cost playing too large a role in model (1) at the expense of the heating degree day term. Clearly, marginal price, which is expected to increase smoothly over time without seasonality, is desirable for accurately modeling consumer behavior here.

Figure 13 provides a scatter plot of $R_1(t)$ versus $Q_1(t)$ for Columbia Gas of Ohio. As in figure 3, the plot character is the last digit of the year of observation with X's drawn through zeros to distinguish 1980 from 1970. Evidently, at any point in time, the relationship between $R_1(t)$ and $Q_1(t)$ is approximately linear. Hence, a TVP model is appropriate as follows:

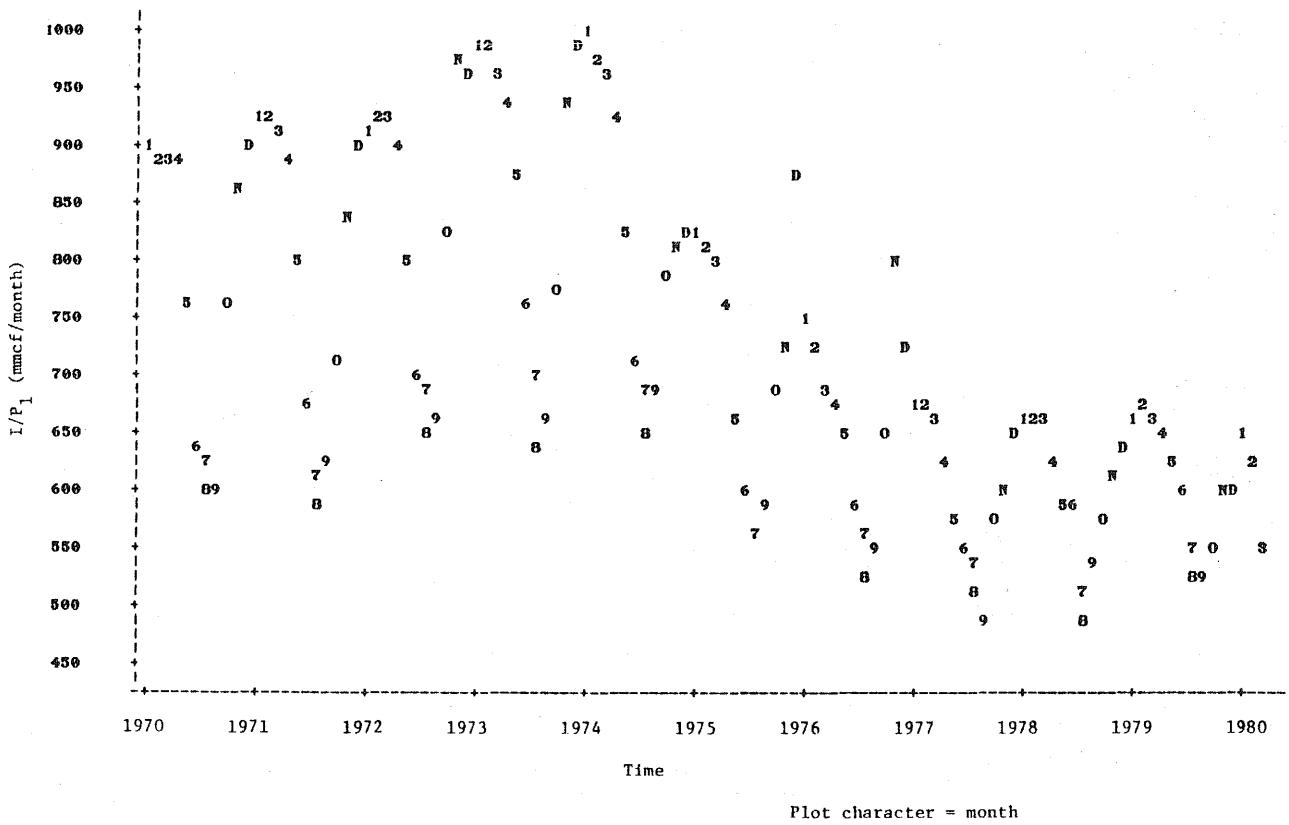


Fig. 12 Time series plot of national personal income divided by natural gas average price: Columbia Gas of Ohio

Marginal Price Model

$$R_1(t) = \lambda_0(t) + \lambda_1(t) Q_1(t) + \varepsilon(t) \quad (6)$$

where $\lambda_0(t)$ = fixed cost; and

$\lambda_1(t)$ = marginal cost (\$/mcf).

Here, $\lambda_1(t)$ yields the aggregate, residential-sector marginal cost needed in model (1) so that $P_1(t) = \lambda_1(t)$.

Values selected for constants and initial values for use in AEP calculations to estimate model (6) included damping factor = 1.0; smoothing constant for means of explanatory variable = 0.01; correction limit = 0.2; $\lambda_0(0) = \lambda_1(0) = 1$; and the number of forward and backward cycles through the data = 30. The large value of 1.0 used for the damping factor is unprecedented and is necessary in order to have AEP keep up with the rapid changes in price. Figure 14 presents the resultant parameter path for $\lambda_1(t)$, the marginal price estimates, for Columbia Gas. In spite of the large damping factor, this is a fairly smooth path showing sharp increases after 1974. Finally, figure 15 shows the $I(t)/P_1(t)$ ratio resulting from use of the estimates from figure 14. Compared to figure 12, which showed this ratio calculated from average cost data, figure 14 shows very little seasonability. The small cyclic behavior from 1975 onward in figure 15 is believed to be an artifact of estimation due to the use of the extremely high damping factor in AEP.

4.2.2 Expansion Model for $Q_1(t)$

Now that marginal price P_1 is available via model (6), it is possible to address the expansion of $Q_0(t)$ and $Q_1(t)$ of model (4) as seen in figures 4 through 11. Attention focuses on $Q_1(t)$, since the space-heating term accounts for over 90% of per capita natural gas consumption. Figures 5, 7, 9, and 11 show that $Q_1(t)$ was constant or increased slightly from 1970 to January 1974; it declined sharply from January 1974 through mid-1977; and finally it remained constant or increased slightly

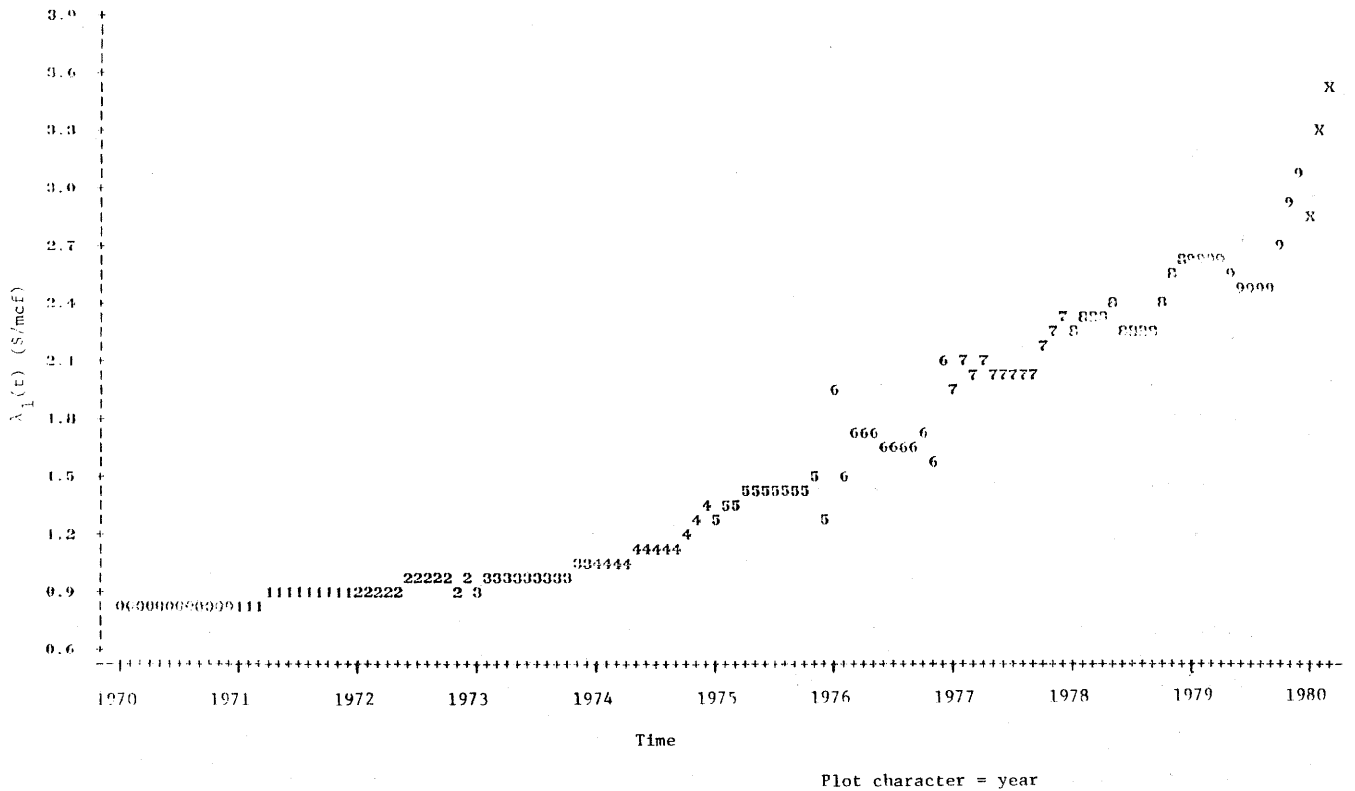


Fig. 14 Parameter path for the slope term, $\alpha_1(t)$, of the marginal price model: Columbia Gas of Ohio

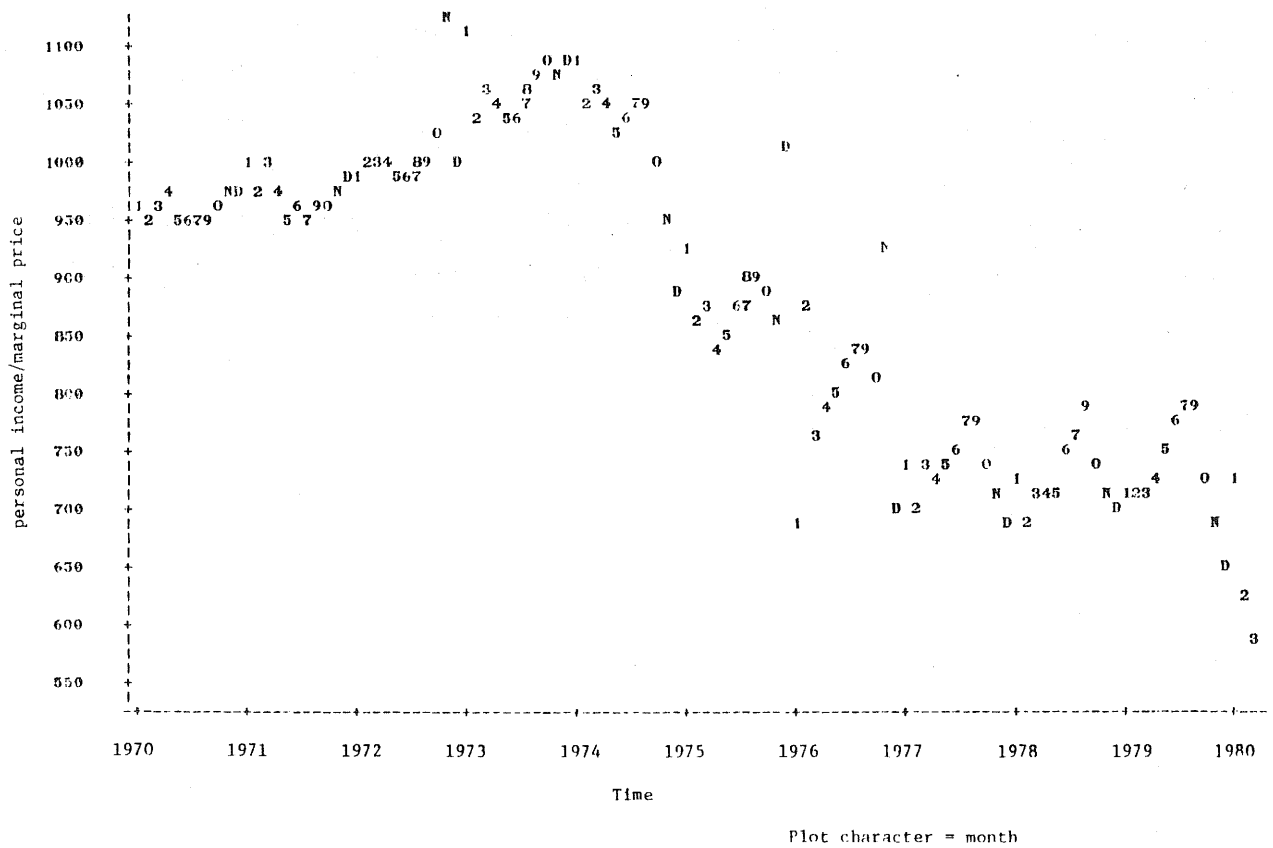


Fig. 15 Time series plot of national personal income divided by natural marginal price: Columbia Gas of Ohio

from mid-1977 through the beginning of 1980. The problem is, then, to find a model using available variables to produce such behavior as in $\hat{\alpha}_1(t)$. Exploratory analysis provided an answer in figure 15: the ratio I/P_1 qualitatively follows the same pattern as $\hat{\alpha}_1(t)$ over time. Therefore, it was decided to use this ratio to expand $\hat{\alpha}_1(t)$ in a model of the form

$$\hat{\alpha}_1(t) = \hat{\gamma}_1 (I/P_1)^{\hat{\gamma}_2} \quad (7)$$

in order to match $\hat{\alpha}_1(t)$ and I/P_1 quantitatively. In raw data form, I/P_1 declines proportionally more than $\hat{\alpha}_1(t)$ so that $\hat{\gamma}_2 < 1$ is expected. Values for $\hat{\gamma}_2$ were found to be roughly 0.5 for Cincinnati Gas and Electric and East Ohio Gas, and 0.7 for the remaining two companies.

4.3 The Final Model

The results of the top-level model and expansion modeling of sections 4.1 and 4.2 provide the final model specification

$$Q_1/N = \beta_0 + \beta_1 (I/P_1)^{\gamma_2} \cdot D + \beta_2 E \cdot D + \beta_3 T \cdot D + \epsilon \quad (8)$$

where γ_2 has the value of 0.5 or 0.7 depending on the distribution utility as given in the previous section. It is not desirable to estimate γ_2 simultaneously with parameters β_0, \dots, β_3 since this would call for a nonlinear estimation procedure. Fortunately, the expansion modeling step provides estimates for γ_2 that are usable here. Note that it is unlikely that the term involving I/P_1 would have been discovered using a conventional modeling procedure. Also note that the term involving T is included here in spite of the evidence from the top-level model that the tax rebate program was ineffective. The inclusion of this term provides a further test as to whether or not the tax rebate program was effective. Both E and T are linked to space-heating uses of natural gas, and so, are interacted with D .

Table 2 summarizes the OLS estimates of model (8) for the four distribution utilities of the case study. The coefficients for the

intercept and income price degree day terms are highly significant; whereas, the emergency coefficient is insignificant, and the tax credit program is insignificant or has the incorrect sign for all but the East Ohio Gas Company. Thus, it appears as if income and price effects in combination with degree days, as expected, explain most of the variation in consumption. The standard errors of estimates in table 2 are seen to be about a tenth of the average gas consumption, and R^2 ranges between 0.969 to 0.986. These are indicators of a good fit to the data. At a significance level of 0.01, the Durbin-Watson test concludes that there is no serial correlation in the estimated models.

Table 3 shows the conservation achieved between January 1974 and January 1978 for all four utilities. Calculations were made here using $D = 900^\circ\text{F}$, roughly the normal winter monthly degree days, to control for temperature differences. The total savings of 16% to 27% compares favorably to similar estimates made from the top-level model in section 4.2.2. Since model (8) of this section is expanded, it is now possible to decompose the total savings of gas per customer into component parts. Thus, table 3 shows the percentage of total savings due to income-price effects versus the tax credit program. Except for East Ohio, 90% to 97% of the savings were due to income-price effects. It is believed that the East Ohio results are not valid, that an alternative transformation of I/P_1 would eliminate the impact of the tax credit dummy variable.

4.4 A Final Diagnostic Check

A goal of the top-down modeling procedure is effectively to guide the analyst in building a complete, constant-parameter model. Systematic variation of TVPs is an indication of missing variables and/or incorrect functional form and in both such cases the model is incomplete. The Durbin-Watson statistics of table 2 indicated no serial correlation in the estimated model (8). This means, possibly, that no systematic variation in parameter paths remains. As a direct check on this conclusion, the TVP version of model (8) was estimated via AEP using the following constants:

TABLE 2

REGRESSION MODEL RESULTS FOR EVALUATION OF
RESIDENTIAL NATURAL GAS CONSERVATION

Company	Parameter Estimates					Statistics				
	Intercept	(Income/Price) ^{Y2} Degree Days	Emergency Degree Days	Tax Credit Degree Days	Number of Observations	F-Value	Standard Error	Average Consumption per Customer	Durbin- Watson Statistic	R ²
Cincinnati Gas & Electric	0.00250*	0.000000810*	0.000000421	-0.000000256	123	2268	0.0012	0.0130	1.59	0.982
Columbia Gas of Ohio	0.00314*	0.000000197*	-0.000000118	0.000000601°	123	2696	0.0012	0.0142	2.07	0.986
Dayton Power & Light	0.00264*	0.000000184*	0.000000518	0.000000005+	123	1266	0.0016	0.0131	1.69	0.969
East Ohio Gas	0.00350*	0.000000844*	0.000000854	-0.000001462*	121	2330	0.0014	0.0168	1.84	0.984

1 * = 0.01 or better significance level for two-sided t test
 + = 0.01 to 0.05 significance
 ° = .05 to 0.10 significance

2 Y2 = 0.5 for Cincinnati Gas and Electric and East Ohio Gas
 Y2 = 0.7 for Columbia Gas and Dayton Power and Light

TABLE 3

ESTIMATED RESIDENTIAL NATURAL GAS CONSERVATION
(mmcf/mo. with Percentages in Parentheses)*

	Per Capita Consumption, <u>January 1974</u>	Per Capita Consumption, <u>January 1978</u>	Total <u>Savings</u>	I/P ₁ <u>Savings</u>	Tax Rebate <u>Savings</u>
Cincinnati Gas & Electric	0.0249 (100)	0.0210 (84)	0.0039 (16)	0.0038 (97)	0.0001 (3)
Columbia Gas of Ohio	0.0268 (100)	0.0206 (77)	0.0062 (23)	0.0058 (94)	0.0003 (5)
Dayton Power & Light	0.0254 (100)	0.0186 (73)	0.0068 (27)	0.0061 (90)	0.0005 (7)
East Ohio Gas	0.0288 (100)	0.0242 (84)	0.0046 (16)	0.0029 (63)	0.0014 (30)

*These estimates use model (8) with estimates from table 2 and data from appendix A. D = 900°F was used in all calculations.

damping factor = 0.06; smoothing constant for means of explanatory variables = 0.01; correction limit = 0.2 $\beta_0(0) = \beta_3(0) = -1$, $\beta_1(0) = \beta_2(0) = 1$; and the number of forward and backwards cycles through the data = 30.

Figures 16 through 19 show the results for $\beta_1(t)$, the coefficient of $(I/P_1)^{\gamma_2} \cdot D$, for each of the four utilities in the case study. As indicated in tables 2 and 3, this is by far the most important term of the model. Indicated on each graph is the percentage decline from the maximum value vertically to the minimum value. These declines are considerably smaller than those of comparable values for the top-level model coefficient $\hat{\alpha}_1(t)$, the coefficient of D.

First, in viewing these figures, all but the East Ohio Gas Company show a similar pattern of decline from 1971 until 1974, followed by an increase throughout the rest of the study period. This pattern is believed merely to be a further transformation of the power law transformation used in the expansion of $\hat{\alpha}_1(t)$ from the top-level model. In other words, the transformation using $\hat{\gamma}_2$ in $(I/P)^{\gamma_2} \cdot D$ was inadequate so that AEP, via the parameter paths in figures 16 through 18, is further transforming the data to improve the fit. A similar argument can be made for figure 19.

Declines from maximum to minimum values are 28%, 26%, 28%, and 22% respectively for figures 5, 7, 9, and 11 of the top-level model. The declines in figures 16 through 19 of 10%, 4%, 14%, and 5% show significant improvement of the final, expanded model versus the top-level model.

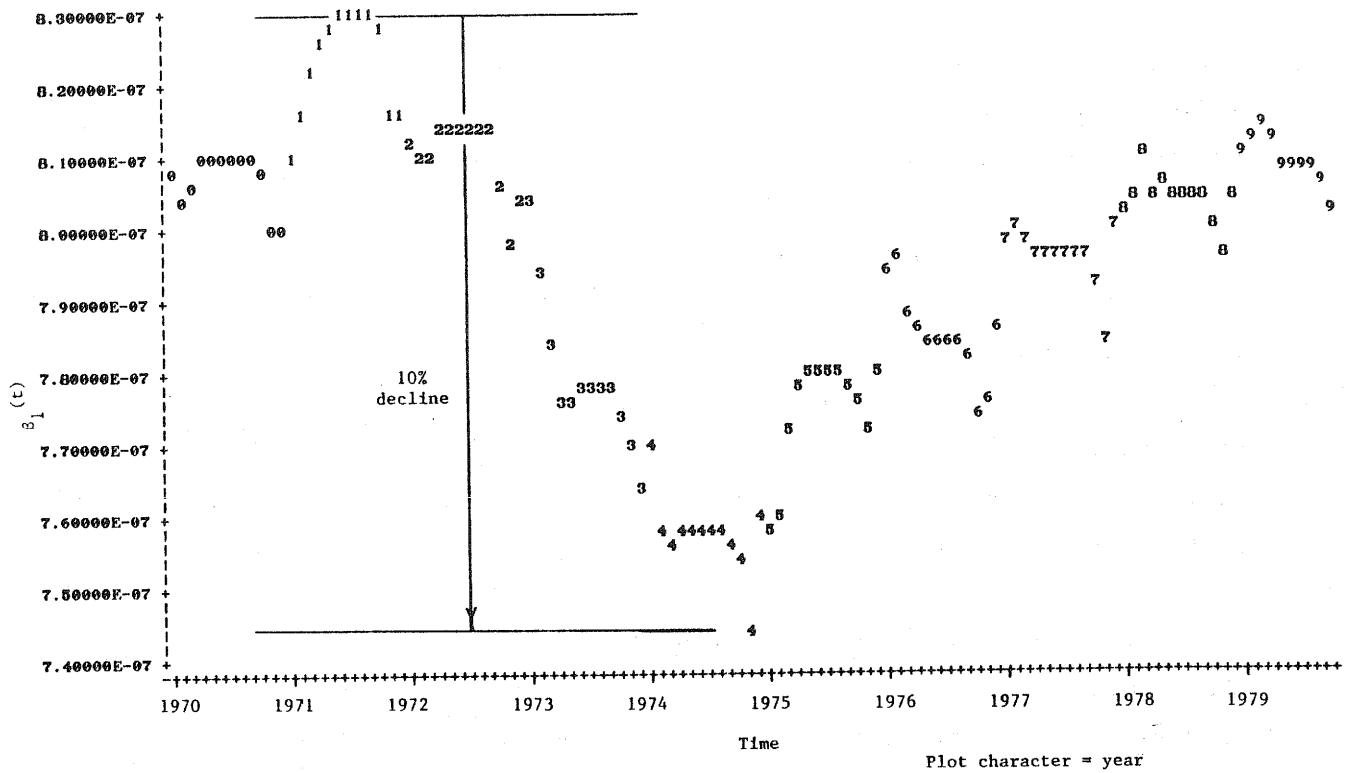


Fig. 16 Parameter path for the $(I/P_1)^{\alpha_2} \cdot D$ term of the expanded model: Cincinnati Gas and Electric

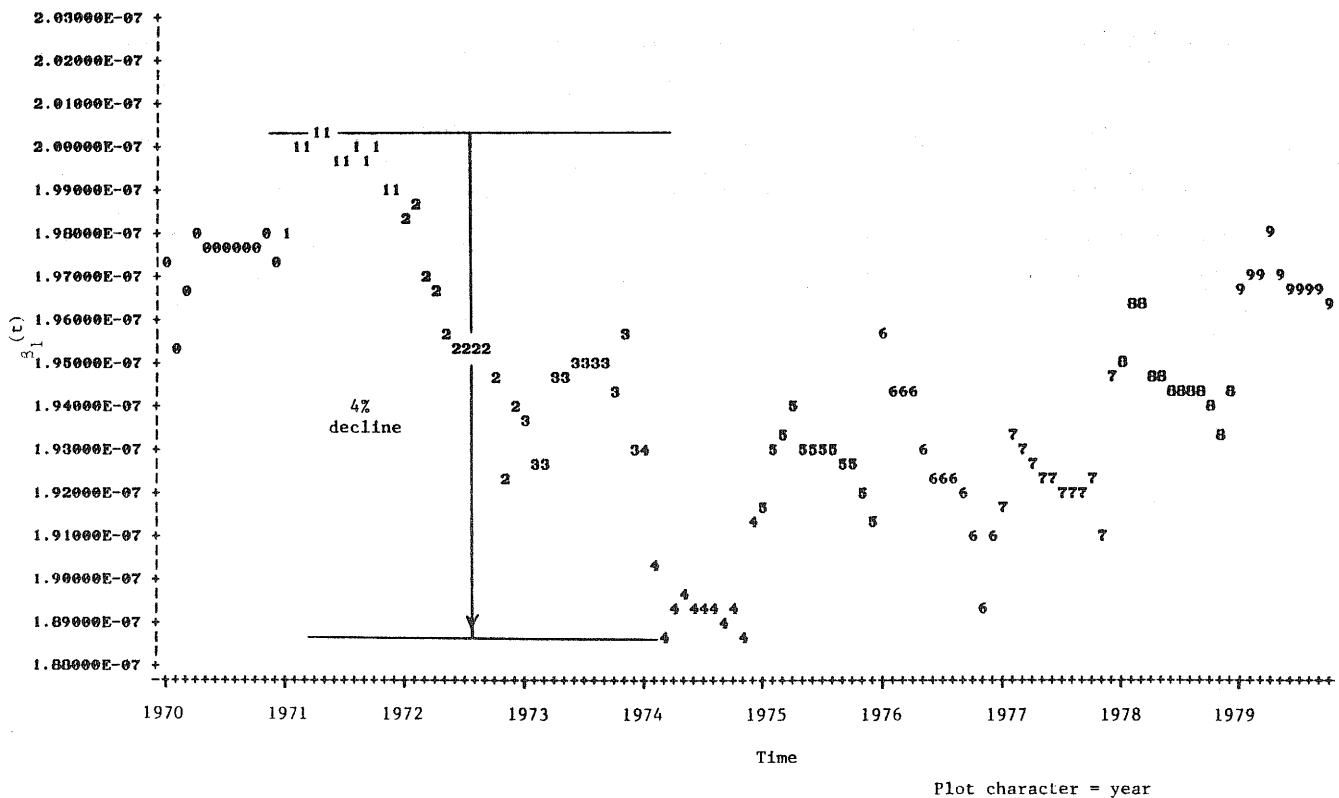


Fig. 17 Parameter path for the $(I/P_1)^{\alpha_2} \cdot D$ term of the expanded model: Columbia Gas of Ohio

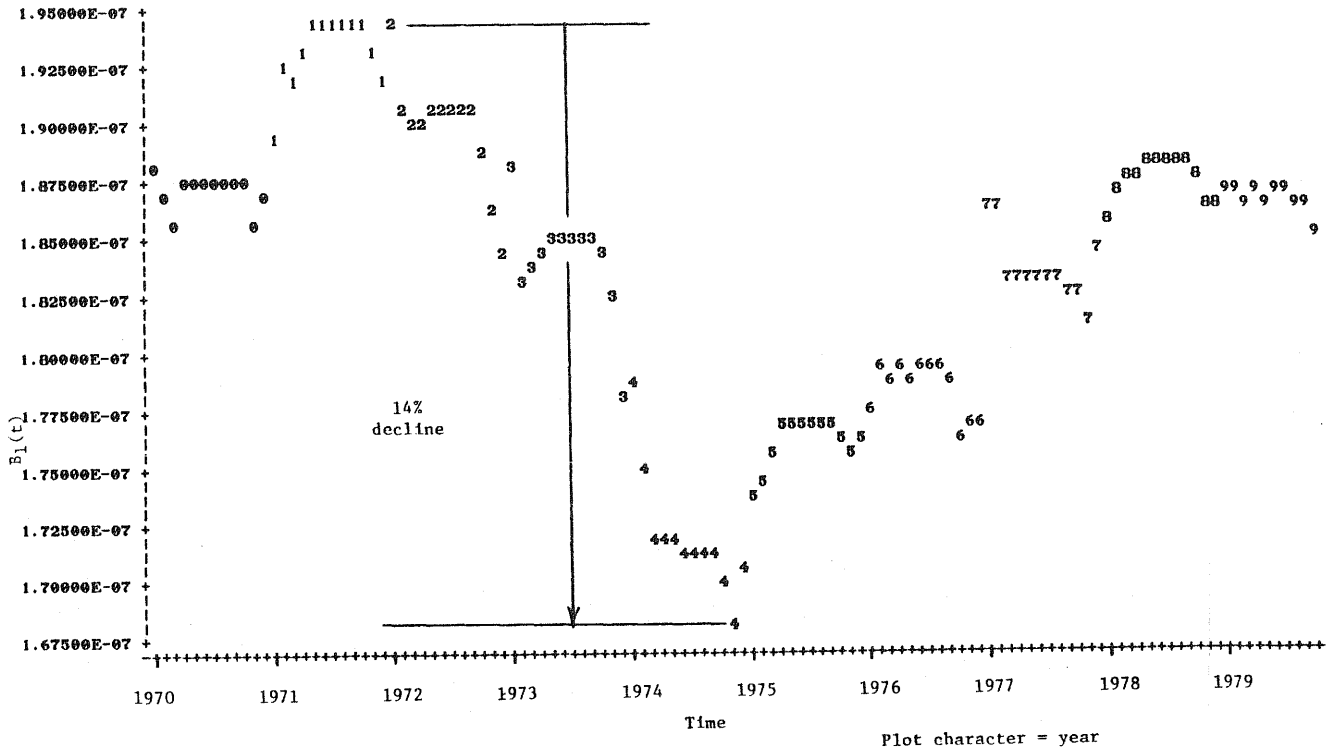


Fig. 18 Parameter path for the $(I/P_1)^{\alpha 2}$. D term of the expanded model: Dayton Power and Light

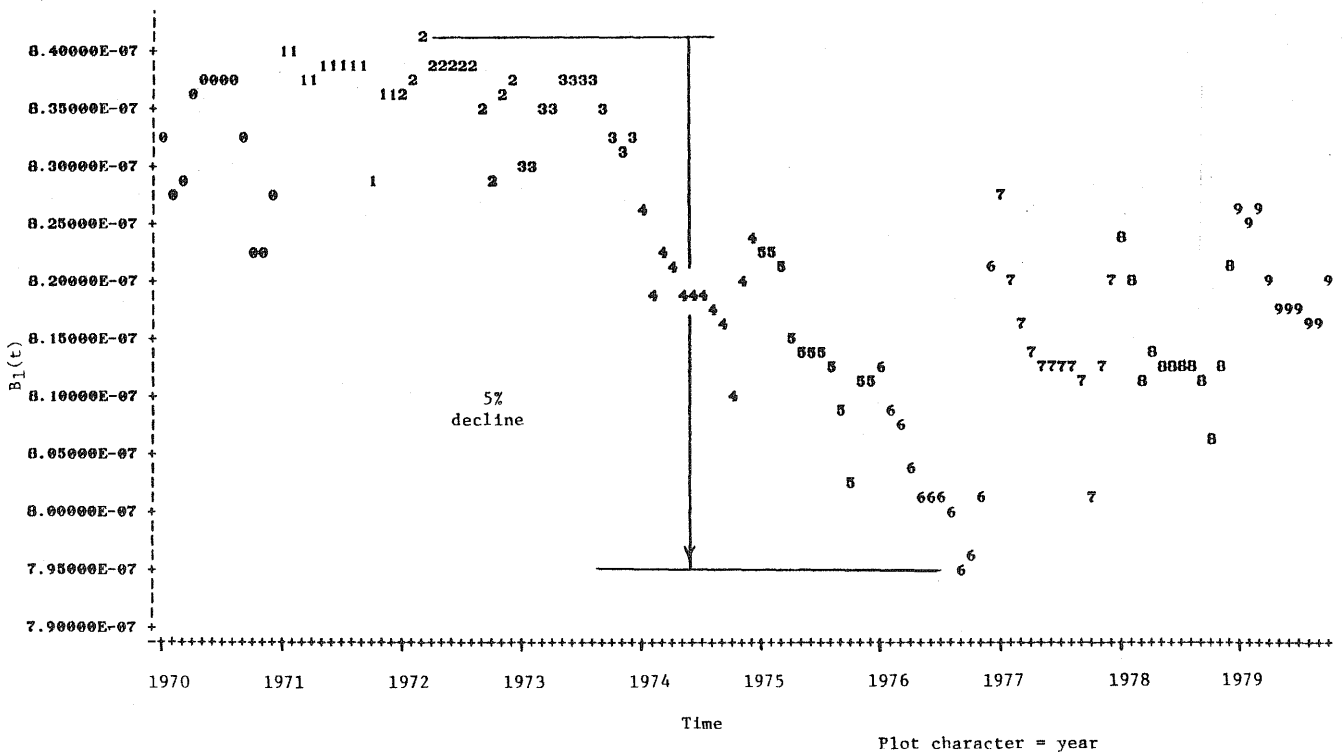


Fig. 19 Parameter path for the $(I/P_1)^{\alpha 2}$. D term of the expanded model: East Ohio Gas

5. Summary

This paper started with a review of long-term trends in space-heat consumption. For a long period until the early 1970s, per capita energy consumption grew strongly as personal income increased faster than the price of space-heat fuel and power. After the October 1973 Arab oil embargo, trends reversed with fuel shortages giving rise to the need for conservation by consumers. Several conservation pressures or incentives were experienced by residential natural gas customers in Ohio including natural gas price hikes starting in 1974, an appeal for emergency conservation during the 1976-77 gas emergency, and the Carter administration's tax credit program that was approved in April of 1977 but probably did not affect consumption, if at all, until late 1977. These conservation pressures were separated nicely over time, and thus, their individual impacts on total conservation should be separable using statistical methods. Intertwined throughout the period of conservation with these conservation measures is the impact of pressures for voluntary conservation. No data on the intensity of voluntarism pressures are available, but even if they were, this component of conservation is not likely to be separable. As a result, conservation impacts due to voluntarism, if any, are mistakenly attributed to the factors for which data are available.

Section 2 of this paper reviewed conventional energy demand models that attempt to explain consumption patterns through price and income effects, and heating (or cooling) degree days. Many researchers notice a limitation in such models that shows up strongly; namely, the behavior of consumers is dynamic and often not accountable through constant-parameter models estimated by ordinary least squares regression. Furthermore, the conventional models failed to produce useful results for the case studies of this paper due to multicollinearity of natural gas price and personal

income. Thus, in section 3, this paper introduced the application of time-varying-parameter models estimated by the adaptive estimation procedure as a means to capture dynamic behavior and to identify a new energy demand model. The combination of time-varying-parameter and constant-parameter models leads to a somewhat new approach to modeling that provides new information to the analyst in constructing models.

Section 4 presented comprehensive case studies on natural gas consumption in the residential sectors of Cincinnati Gas and Electric, Columbia Gas of Ohio, Inc., Dayton Power and Light, and East Ohio Gas. Top-level time-varying-parameter models of per customer consumption, using only heating degree days as an explanatory variable, showed from 14% to 20% conservation over the period from January 1974 to January 1978. The preliminary model provided strong evidence that the tax credit program had little impact on total conservation. Perhaps, this program converted temporary behavior, such as reduced thermostat settings, into more permanent conservation measures such as the implementation of weatherization materials.

Section 4 also presented expanded consumption models that included additional explanatory variables. These variables included personal income, marginal natural gas price estimated via a time-varying-parameter model, and dummy variables for the gas emergency and the tax rebate program. Almost all of the total conservation was found to be attributable to income and price effects.

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Appendix A

Listing, Documentation, and Discussion of Raw Data

Tables A-1 through A-4 list data on monthly natural gas consumption in the residential sectors of Cincinnati Gas and Electric Company, Columbia Gas of Ohio, Inc., Dayton Power and Light Company, and the East Ohio Gas Company respectively. The text of this appendix documents and discusses the variables of these tables.

Year and month are self-explanatory. Residential natural gas consumption, labeled "Gas" in the tables, and number of customers were obtained directly from the distribution utilities and from the Public Utilities Commission of Ohio. Consideration must be given to three aspects of these data: (1) they are aggregated from meter reading data, (2) half of each month's data from Columbia Gas are estimated, and (3) many customers are on a budget billing plan. Each aspect is discussed in turn.

First, the so-called "monthly" consumption datum for a month is actually the sum of all meter readings taken within, and a few days after, the month. The total set of customers is partitioned into 21 billing units with 21 different meter reading days in a month. Thus, consumption reported for a given month actually spans a period of two months, the given month and the previous month. Fisch et al. [A.1] have shown that a two-month moving average of heating degree days provides good estimates of such data. The moving average approach compares favorably with the alternative of modeling each billing unit separately, using the exactly corresponding degree days, and then aggregating to the equivalent monthly total. This paper uses a two-month moving average for degree days, but not for any other variables in the model. The income and price variables change slowly on a month-to-month basis so that averaging provides little benefit for these variables.

The second issue is that residential customers of Columbia Gas have meters read every other month. Consider a specific customer and two consecutive months A and B. Suppose that month A does not have a meter reading for the customer, then month A's consumption is estimated from a model determined from recent data on the customer's behavior and measured

TABLE A-1

CINCINNATI GAS AND ELECTRIC:
DATA SET ON RESIDENTIAL NATURAL GAS
CONSUMPTION AND ITS DETERMINANTS

Year	Month	Gas (mmcf)	Number of Customers	Revenue (\$10 ³)	Marginal Price	National Income (\$10 ⁹)	Consumer Price Index	Degree Days (°F)
70	1	8145	260067	7089	0.85415	770.8	113.3	1162
70	2	7626	260544	6691	0.86095	781.5	113.9	1110
70	3	6337	260689	5597	0.85783	787.6	114.5	876
70	4	4489	260585	4044	0.86862	806.0	115.2	531
70	5	1844	260038	1800	0.87979	799.7	115.7	178
70	6	1036	259154	1101	0.88730	798.2	116.3	47
70	7	878	258022	963	0.89227	803.3	116.7	3
70	8	804	258823	899	0.89580	806.4	116.9	2
70	9	815	259077	910	0.89829	811.9	117.5	14
70	10	1284	260095	1312	0.89515	813.6	118.1	120
70	11	2862	261448	2663	0.87669	815.7	118.5	402
70	12	5410	262478	4804	0.84790	820.9	119.1	710
71	1	7751	263276	6781	0.86089	830.0	119.2	981
71	2	7868	263926	6919	0.85615	833.2	119.4	1007
71	3	6695	264187	5955	0.87244	839.7	119.8	820
71	4	4711	263943	4377	0.89969	844.4	120.2	555
71	5	2674	263409	2772	0.95090	850.0	120.8	254
71	6	1430	262794	1600	0.96542	870.1	121.5	79
71	7	939	262194	1116	0.97040	859.2	121.8	0
71	8	876	261982	1052	0.97275	867.6	122.2	0
71	9	896	262357	1075	0.97504	871.5	122.4	16
71	10	1096	263232	1265	0.97328	874.8	122.6	80
71	11	2419	264287	2538	0.97084	879.4	122.6	389
71	12	5455	265698	5436	0.95590	890.4	123.1	690
72	1	6455	266727	6421	0.97193	898.9	123.2	902
72	2	7687	267882	7598	0.95355	908.5	123.8	1049
72	3	6688	267789	6653	0.97599	913.6	124.0	888
72	4	4817	267500	4870	0.96904	919.4	124.3	563
72	5	2462	266774	2618	0.97891	924.0	124.7	245
72	6	1357	266161	1545	0.98376	922.9	125.0	98
72	7	1082	265601	1272	0.98573	932.9	125.5	41
72	8	841	265330	1037	0.98743	940.0	125.7	3
72	9	927	265330	1123	0.98875	946.8	126.2	14
72	10	1543	266223	1753	0.99492	967.0	126.6	214
72	11	3986	267583	4179	0.99693	977.6	126.9	569
72	12	6730	268824	6850	0.97885	983.6	127.3	812
73	1	7340	269580	7442	0.99332	989.1	127.7	958
73	2	6786	270052	6868	0.97212	997.1	128.6	957
73	3	5474	270271	5618	0.99699	1000.3	129.8	800
73	4	4050	270288	4247	0.99760	1011.6	130.7	568
73	5	2949	269825	3173	1.00351	1018.7	131.5	320
73	6	1543	269350	1847	1.025	1026.6	132.4	105
73	7	887	268989	1166	1.034	1035.0	132.7	2
73	8	834	268801	1111	1.039	1047.3	135.1	1
73	9	900	269056	1188	1.044	1058.5	135.5	14
73	10	1165	270081	1485	1.052	1090.8	136.6	122
73	11	3270	271440	3728	1.064	1100.0	137.6	400
73	12	5316	272182	5906	1.06533	1107.1	138.5	734
74	1	7293	272531	8088	1.08532	1107.0	139.7	933
74	2	5886	272799	6690	1.10010	1113.4	141.5	882
74	3	5326	272734	6178	1.11910	1117.1	143.1	727
74	4	4018	272291	4826	1.13987	1125.2	144.0	462
74	5	2318	271368	3064	1.18247	1135.2	145.6	247
74	6	1291	270713	1801	1.18359	1143.5	147.1	96
74	7	971	270289	1400	1.17906	1159.5	148.3	16
74	8	878	269915	1297	1.17794	1167.2	150.2	0
74	9	1055	269951	1541	1.18568	1178.0	151.9	74
74	10	2141	270448	2965	1.22620	1185.0	153.2	259
74	11	3078	270899	4196	1.26264	1184.5	154.3	493
74	12	6465	271391	8894	1.38320	1191.0	155.4	778

TABLE A-1
(continued)

Year	Month	Gas (mmcf)	Number of Customers	Revenue (\$10 ³)	Marginal Price	National Income (\$10 ⁹)	Consumer Price Index	Degree Days (°F)
75	1	6516	271596	948	1.35731	1191.1	156.1	9230
75	2	6197	271672	864	1.35015	1193.4	157.2	8693
75	3	6096	271571	776	1.46138	1195.7	157.8	8949
75	4	4760	271192	607	1.53563	1203.1	158.6	7556
75	5	2367	270460	243	1.55073	1214.3	159.3	4028
75	6	1053	269731	30	1.53531	1244.1	160.6	1906
75	7	800	269203	8	1.51755	1238.9	162.3	1477
75	8	806	268775	3	1.46946	1255.9	162.8	1324
75	9	954	268687	71	1.47762	1270.9	163.3	1704
75	10	1786	269286	193	1.49553	1290.8	164.6	2981
75	11	2626	269693	359	1.51467	1300.2	165.6	4273
75	12	5377	270074	661	1.54533	1308.2	166.3	8539
76	1	7612	270423	1008	1.64813	1320.8	166.7	12420
76	2	6592	270456	919	1.63323	1331.4	167.1	11108
76	3	3988	270256	614	1.64585	1341.9	167.5	6856
76	4	3233	269808	440	1.62220	1352.5	168.2	5531
76	5	2170	269395	249	1.60812	1362.9	169.2	3760
76	6	1231	268816	91	1.59890	1370.4	170.1	2232
76	7	842	268221	3	1.60984	1380.8	171.1	1639
76	8	847	267804	2	1.60950	1385.5	171.9	1678
76	9	930	267863	38	1.64474	1391.7	172.6	1933
76	10	1730	268365	285	1.79912	1414.2	173.3	3744
76	11	4687	268940	696	2.12238	1432.1	173.8	9958
76	12	6810	269248	1025	2.26020	1450.2	174.3	15365
77	1	8801	269466	1399	2.29157	1454.3	175.3	20441
77	2	7951	269469	1310	2.30107	1477.0	177.1	18694
77	3	4708	269203	776	2.38664	1499.1	178.2	11590
77	4	2837	268725	424	2.39369	1510.1	179.6	7250
77	5	1645	267886	172	2.39911	1517.3	180.6	4413
77	6	964	266875	52	2.41097	1524.3	181.8	2815
77	7	865	266111	18	2.41676	1539.2	182.6	2577
77	8	787	265514	1	2.42629	1540.7	183.3	2412
77	9	808	265365	17	2.45036	1556.9	184.0	2540
77	10	1530	266199	212	2.51410	1577.0	184.5	4492
77	11	2617	266692	489	2.56698	1592.7	185.4	7296
77	12	5944	267203	852	2.50149	1609.2	186.1	15564
78	1	7419	267352	1279	2.50997	1615.5	187.2	19121
78	2	8020	267494	1372	2.52217	1625.0	188.4	20707
78	3	6690	267409	1092	2.60639	1646.3	189.8	17715
78	4	3460	266626	613	2.63284	1669.4	191.5	9661
78	5	2301	265997	277	2.62856	1682.1	193.3	6591
78	6	1204	265369	109	2.65223	1695.7	195.3	3810
78	7	804	264557	5	2.65299	1730.0	196.7	2698
78	8	701	264128	0	2.65372	1741.3	197.8	2426
78	9	834	262571	11	2.65676	1756.1	199.3	2790
78	10	1464	264656	201	2.66910	1781.0	200.9	4504
78	11	2892	265221	467	2.68238	1801.4	202.0	8340
78	12	4917	265865	722	2.68909	1826.8	202.9	13789
79	1	7126	266219	1120	2.65439	1834.3	204.7	19616
79	2	7768	266429	1282	2.68488	1851.4	207.1	21302
79	3	5703	266327	890	2.66813	1872.1	209.1	15827
79	4	3424	265822	494	2.66741	1880.7	211.5	9707
79	5	1984	265097	302	2.67090	1891.6	214.1	5876
79	6	1159	264366	97	2.67137	1905.1	216.6	3675
79	7	855	263765	8	2.66889	1913.2	218.9	2849
79	8	760	263466	8	2.65296	1946.5	221.1	2532
79	9	843	263604	37	2.69924	1960.1	223.4	2991
79	10	1516	264336	218	2.88955	1981.2	225.4	5426
79	11	3095	265091	496	3.14004	2005.5	227.5	10545
79	12	4845	266084	752	3.19597	2028.3	229.9	16169
80	1	6008	266752	984	3.39831	2046.5	233.2	20686
80	2	6667	267156	1131	3.45069	2055.6	236.4	23637
80	3	6051	267423	998	3.42685	2069.6	239.8	21594
80	4	242.5	14077

TABLE A-2

COLUMBIA GAS OF OHIO:
DATA SET ON RESIDENTIAL NATURAL GAS
CONSUMPTION AND ITS DETERMINANTS

Year	Month	Gas (mmcf)	Number of Customers	Revenue (\$10 ³)	Marginal Price	National Income (\$10 ⁹)	Consumer Price Index	Degree Days (°F)
70	1	31033	915609	26735	0.81437	778.8	113.3	1272
70	2	27301	916375	23955	0.82218	781.5	113.9	1193
70	3	24851	916469	21998	0.81916	787.6	114.5	939
70	4	18418	915529	16824	0.82730	806.0	115.2	613
70	5	8007	911991	8441	0.83842	799.7	115.7	242
70	6	4367	909352	5468	0.84563	798.2	116.3	66
70	7	4052	908094	5224	0.85004	803.3	116.7	12
70	8	3516	908041	4771	0.85215	806.4	116.9	5
70	9	3556	910405	4835	0.85453	811.9	117.5	29
70	10	6963	918190	7486	0.84122	813.6	118.1	178
70	11	14492	925796	13764	0.83194	815.7	118.5	436
70	12	21190	930947	19461	0.84971	820.9	119.1	809
71	1	29278	932212	26155	0.82922	830.0	119.2	1100
71	2	30411	933347	27264	0.85227	833.2	119.4	1103
71	3	24805	933832	22681	0.84941	839.7	119.8	911
71	4	19112	932118	18139	0.86382	844.4	120.2	673
71	5	11036	930216	11760	0.89258	850.0	120.8	353
71	6	5103	927337	6589	0.90349	870.1	121.5	117
71	7	3818	925734	5409	0.90716	859.2	121.8	4
71	8	3402	926277	5015	0.90808	867.6	122.2	4
71	9	3647	929641	5124	0.90331	871.5	122.4	29
71	10	6222	936709	7627	0.91034	874.8	122.6	117
71	11	13007	945019	13702	0.90726	879.4	122.6	457
71	12	21292	950757	21167	0.90371	890.4	123.1	774
72	1	25808	952768	25384	0.91452	898.9	123.2	974
72	2	29933	954541	29227	0.91966	908.5	123.8	1105
72	3	24562	955390	24390	0.91816	913.6	124.0	959
72	4	18830	954622	19259	0.92158	919.4	124.3	671
72	5	9122	952148	10537	0.93358	924.0	124.7	314
72	6	5232	949492	6959	0.93937	922.9	125.0	124
72	7	4954	949235	6685	0.94129	932.9	125.5	62
72	8	4130	949714	5947	0.94390	940.0	125.7	20
72	9	4190	954875	6028	0.94629	946.8	126.2	48
72	10	8930	958765	10483	0.94585	967.0	126.6	275
72	11	16128	965903	16087	0.86855	977.6	126.9	600
72	12	24028	970857	24465	0.98085	983.6	127.3	808
73	1	27516	972401	27713	0.89360	989.1	127.7	965
73	2	26192	973807	26370	0.96021	997.4	128.6	988
73	3	20508	973955	21403	0.94921	1003.3	129.8	689
73	4	16261	973622	17645	0.96271	1011.6	130.7	436
73	5	10989	970817	12885	0.97864	1018.7	131.5	306
73	6	5847	967537	7908	0.98415	1026.6	132.4	92
73	7	4381	966705	6495	0.98776	1035.6	132.7	0
73	8	3282	966444	5367	0.98742	1047.3	135.1	2
73	9	3632	965511	5755	0.98919	1058.5	135.5	19
73	10	5716	968962	8077	1.00392	1090.8	136.6	127
73	11	15037	973888	17559	1.01931	1100.0	137.6	404
73	12	20386	976935	22994	1.01685	1107.1	138.5	776
74	1	27700	977750	30504	1.02277	1107.0	139.7	970
74	2	24084	977791	27519	1.06403	1113.4	141.5	959
74	3	20349	977324	23693	1.04692	1117.1	143.1	704
74	4	15962	975101	19435	1.07448	1125.2	144.0	480
74	5	9510	972026	13109	1.10663	1135.2	145.6	255
74	6	4958	968151	7946	1.10766	1143.5	147.1	105
74	7	4228	965461	7142	1.10331	1159.5	148.3	16
74	8	3653	961736	6501	1.10843	1167.2	150.2	0
74	9	4424	965809	7555	1.11821	1178.0	151.9	65
74	10	9443	970579	14233	1.18667	1185.0	153.2	252
74	11	13612	973576	19888	1.24904	1184.5	154.3	492
74	12	22763	975744	32841	1.34987	1191.0	155.4	782

TABLE A-2
(continued)

Year	Month	Gas (mmcf)	Number of Customers	Revenue (\$10 ³)	Marginal Price	National Income (\$10 ⁹)	Consumer Price Index	Degree Days (°F)
75	1	25279	975188	36311	1.29432	1191.1	156.1	977
75	2	25121	974895	36707	1.37547	1193.4	157.2	939
75	3	22326	973908	33619	1.36579	1195.7	157.8	864
75	4	18558	972051	29393	1.42920	1203.1	158.6	696
75	5	8307	968193	15099	1.43332	1214.3	159.3	308
75	6	4668	963749	9690	1.42506	1244.1	160.6	46
75	7	3800	961822	8279	1.41356	1238.9	162.3	9
75	8	3311	960882	7444	1.40209	1255.9	162.8	0
75	9	4025	963653	8630	1.40525	1270.9	163.3	55
75	10	7928	967774	14953	1.45587	1290.8	164.6	216
75	11	11383	970772	20396	1.49926	1300.2	165.6	421
75	12	20253	973661	30405	1.28723	1308.2	166.3	747
76	1	28092	973999	49211	1.93084	1320.8	166.7	1118
76	2	24401	973467	44737	1.52537	1331.4	167.1	1027
76	3	16632	971487	32322	1.76176	1341.9	167.5	681
76	4	13396	968529	26770	1.73078	1352.5	168.2	505
76	5	8339	965561	17424	1.69556	1362.9	169.2	335
76	6	4777	960988	11063	1.66935	1370.4	170.1	117
76	7	3750	958208	9234	1.65261	1380.8	171.1	3
76	8	3341	957097	8723	1.65258	1385.5	171.9	17
76	9	3959	958923	9921	1.66139	1391.7	172.6	80
76	10	8813	963638	19077	1.73311	1414.2	173.3	340
76	11	17940	968559	31873	1.55296	1432.1	173.8	740
76	12	24575	970618	49169	2.08383	1450.2	174.3	1099
77	1	30958	970471	66698	1.97239	1454.3	175.3	1462
77	2	29285	970248	63512	2.10996	1477.0	177.1	1377
77	3	19021	969074	42997	2.04601	1499.1	178.2	852
77	4	11687	966382	28995	2.06703	1510.1	179.6	476
77	5	6343	960324	16674	2.05370	1517.3	180.6	214
77	6	4434	954783	12216	2.02085	1524.3	181.8	78
77	7	3562	951273	10251	1.99591	1539.9	182.6	33
77	8	3137	949346	9526	1.98976	1540.7	183.3	9
77	9	3066	950853	9784	2.00588	1556.9	184.0	27
77	10	7460	957318	20668	2.14169	1577.0	184.5	215
77	11	10662	961531	28252	2.23961	1592.7	185.4	494
77	12	21530	965316	53727	2.33553	1609.2	186.1	843
78	1	26993	965532	65915	2.24240	1615.5	187.2	1256
78	2	29259	965339	71285	2.34792	1625.0	188.4	1383
78	3	24320	963355	60733	2.30785	1646.3	189.8	1142
78	4	13901	959390	37099	2.36106	1669.4	191.5	681
78	5	9154	955812	25985	2.36652	1682.1	193.3	324
78	6	4152	950415	11880	2.25252	1695.7	195.3	123
78	7	3574	946050	12123	2.27580	1730.0	196.7	12
78	8	3042	943695	10886	2.28574	1741.3	197.8	0
78	9	3123	944643	10177	2.24728	1756.1	199.3	19
78	10	6986	951301	21715	2.39812	1781.0	200.9	225
78	11	11560	956440	33968	2.54521	1801.4	202.0	511
78	12	18157	960680	51545	2.61663	1826.8	202.9	777
79	1	26093	961475	72302	2.59295	1834.3	204.7	1145
79	2	27715	961711	76523	2.61469	1851.4	207.1	1308
79	3	20585	959696	58459	2.63223	1872.1	209.1	954
79	4	14342	956267	41464	2.58753	1880.7	211.5	543
79	5	7749	951582	23491	2.53417	1891.6	214.1	312
79	6	4615	946110	14542	2.46225	1905.1	216.6	102
79	7	3557	943636	12488	2.45755	1933.2	218.9	15
79	8	3123	942243	11452	2.45675	1946.5	221.1	14
79	9	3432	945045	12731	2.48342	1960.1	223.4	50
79	10	7049	953017	25127	2.71103	1981.2	225.4	230
79	11	12167	960558	41078	2.94223	2005.5	227.5	594
79	12	17498	965916	58820	3.09997	2028.3	229.9	776
80	1	23384	967620	74011	2.83854	2046.5	233.2	1010
80	2	24994	969111	82888	3.32137	2055.6	236.4	1124
80	3	21292	968867	79540	3.55254	2069.6	239.8	1002
80	4	.	.	53502	.	.	242.5	.

TABLE A-3

DAYTON POWER AND LIGHT:
DATA SET ON RESIDENTIAL NATURAL GAS
CONSUMPTION AND ITS DETERMINANTS

Year	Month	Gas (mmcf)	Number of Customers	Revenue (\$10 ³)	Marginal Price	National Income (\$10 ⁹)	Consumer Price Index	Degree Days (°F)
70	1	8140	235883	6116	0.70493	778.8	113.3	1242
70	2	7027	236114	5385	0.76672	781.5	113.9	1182
70	3	5846	236255	4542	0.71933	787.6	114.5	919
70	4	4661	236261	3683	0.75061	806.0	115.2	590
70	5	2603	235969	1785	0.76430	799.7	115.7	226
70	6	1099	235780	1116	0.77468	798.2	116.3	63
70	7	837	235745	912	0.77936	803.3	116.7	8
70	8	779	235899	865	0.78199	806.4	116.9	3
70	9	775	236232	864	0.78356	811.9	117.5	24
70	10	1477	236827	1394	0.77945	813.6	118.1	163
70	11	3018	237763	2700	0.80680	815.7	118.5	482
70	12	5178	238824	4671	0.87466	820.9	119.1	796
71	1	7636	239603	6744	0.81397	830.0	119.2	1086
71	2	7551	240032	6710	0.89991	833.2	119.4	1120
71	3	5869	240151	5341	0.83831	839.7	119.8	906
71	4	4593	240098	4295	0.88777	844.4	120.2	620
71	5	2656	239734	2815	0.93542	850.0	120.8	298
71	6	1417	239578	1669	0.94684	870.1	121.5	97
71	7	793	239392	1066	0.94894	859.2	121.8	1
71	8	761	239553	1034	0.94973	867.6	122.2	0
71	9	827	240011	1101	0.95111	871.5	122.4	25
71	10	1332	240778	1589	0.95305	874.8	122.4	74
71	11	2396	241876	2571	0.94634	879.4	122.6	379
71	12	5022	243479	4682	0.83676	890.4	123.1	734
72	1	6622	244184	6440	1.01359	898.9	123.2	981
72	2	6937	244835	6716	0.83194	908.5	123.8	1107
72	3	5857	245233	5760	1.00017	913.6	124.0	964
72	4	4629	245138	4669	0.92138	919.4	124.3	654
72	5	2465	244697	2697	0.95323	924.0	124.7	284
72	6	1350	244234	1643	0.96218	922.9	125.0	99
72	7	1109	243991	1405	0.96488	932.9	125.5	41
72	8	817	244190	1118	0.96503	940.0	125.7	11
72	9	837	244498	1139	0.96548	946.8	126.2	35
72	10	1525	245044	1814	0.96846	967.0	126.6	253
72	11	3711	246487	3842	0.94485	977.6	126.9	607
72	12	6161	247901	5586	0.78325	983.6	127.3	869
73	1	7119	248524	6947	1.09090	989.1	127.7	1003
73	2	6199	248870	6121	0.80963	997.4	128.6	990
73	3	4999	249008	5035	0.99960	1003.3	129.8	723
73	4	3933	248649	4102	0.95121	1011.6	130.7	473
73	5	2712	248037	2960	0.96346	1018.7	131.5	324
73	6	1521	247572	1823	0.96795	1026.6	132.4	102
73	7	864	247198	1177	0.96755	1035.6	132.7	0
73	8	741	246809	1045	0.96468	1047.3	135.1	1
73	9	812	246642	1119	0.96429	1058.5	135.5	17
73	10	1125	246735	1468	0.97514	1090.8	136.6	129
73	11	2777	247076	3091	0.98473	1100.0	137.6	410
73	12	4633	247473	4905	0.98212	1107.1	138.5	819
74	1	7222	247696	7347	0.95407	1107.0	139.7	1025
74	2	5541	247722	5870	1.02099	1113.4	141.5	995
74	3	4750	247617	5101	0.99031	1117.1	143.1	839
74	4	3894	247356	4313	1.01858	1125.2	144.0	535
74	5	2240	245777	2797	1.05826	1135.2	145.6	289
74	6	1236	245129	1704	1.06123	1143.5	147.1	120
74	7	907	244555	1350	1.06152	1159.5	148.3	21
74	8	746	244344	1162	1.05814	1167.2	150.2	0
74	9	924	244289	1402	1.06773	1178.0	151.9	82
74	10	1879	244559	2614	1.12338	1185.0	153.2	274
74	11	2743	245069	3749	1.18864	1184.5	154.3	499
74	12	5279	245608	7174	1.31573	1191.0	155.4	801

TABLE A-3
(continued)

Year	Month	Gas (MMCF)	Number of Customers	Revenue (10 ³)	Marginal Price	National Income (\$10 ⁹)	CP Index	Degree Days (°F)
75	1	6488	245865	8714	1.22606	1191.1	156.1	983
75	2	5659	245930	7725	1.31584	1193.4	157.2	923
75	3	5273	245927	7472	1.32786	1195.7	157.8	852
75	4	4421	245635	6641	1.40463	1203.1	158.6	683
75	5	2153	245142	3599	1.42000	1214.3	159.3	307
75	6	979	244609	1932	1.42396	1244.1	160.6	53
75	7	768	244238	1610	1.42140	1238.9	162.3	12
75	8	735	243930	1561	1.41966	1255.9	162.8	1
75	9	868	243884	1813	1.43259	1270.9	163.3	65
75	10	1673	244070	3086	1.47075	1290.8	164.6	211
75	11	2345	244459	4194	1.51867	1300.2	165.6	392
75	12	4306	244939	7433	1.57926	1308.8	166.3	725
76	1	6074	245234	10448	1.65850	1320.8	166.7	1198
76	2	6007	245320	10602	1.66266	1331.4	167.1	1010
76	3	3746	245242	6870	1.66707	1341.9	167.5	665
76	4	3106	244583	5758	1.65418	1352.5	168.2	487
76	5	2090	244337	3935	1.61655	1362.9	169.2	303
76	6	1242	243827	2534	1.59895	1370.4	170.1	103
76	7	789	243438	1782	1.58558	1380.8	171.1	2
76	8	716	243190	1687	1.58314	1385.5	171.9	15
76	9	794	243044	1868	1.59443	1391.7	172.6	79
76	10	1575	243345	3241	1.63440	1414.2	173.3	329
76	11	4184	244032	8020	1.80268	1432.1	173.8	726
76	12	5766	244481	10628	1.67919	1450.2	174.3	1053
77	1	7542	244559	15734	2.40144	1454.3	175.3	1371
77	2	7706	244584	15953	1.43233	1477.0	177.1	1299
77	3	3480	244445	7727	2.01158	1499.1	178.2	834
77	4	2861	244118	6455	1.97199	1510.1	179.6	465
77	5	1513	243906	3646	1.93181	1517.3	180.6	193
77	6	833	243582	2229	1.89585	1524.3	181.8	50
77	7	715	243048	1994	1.87614	1539.2	182.6	18
77	8	638	242782	1834	1.86332	1540.7	183.3	4
77	9	737	242512	2139	1.88204	1556.9	184.0	26
77	10	1480	242407	4089	2.03399	1577.0	184.5	206
77	11	2393	242758	6486	2.25033	1592.7	185.4	486
77	12	4845	243379	12403	2.41366	1609.2	186.1	857
78	1	6414	243630	16999	2.30438	1615.5	187.2	1268
78	2	6848	243657	17129	2.41709	1625.0	188.4	1384
78	3	5766	243695	14575	2.31653	1646.3	189.8	1133
78	4	3495	243309	9320	2.38329	1669.4	191.5	661
78	5	2349	242840	6499	2.36127	1682.1	193.3	300
78	6	1062	242336	3292	2.31825	1695.7	195.3	110
78	7	734	241820	2465	2.28588	1730.0	196.7	8
78	8	654	241333	2282	2.26996	1741.3	197.8	6
78	9	688	241139	2400	2.27015	1756.1	199.3	29
78	10	1427	241107	4253	2.31496	1781.0	200.9	222
78	11	2532	241322	7055	2.40438	1801.4	202.0	497
78	12	4087	241601	10994	2.47824	1826.8	202.9	770
79	1	5990	242042	15751	2.46916	1834.3	204.7	1157
79	2	6745	242508	17619	2.47795	1851.4	207.1	1327
79	3	4921	242499	13082	2.46389	1872.1	209.1	958
79	4	3264	242122	8918	2.44790	1880.7	211.5	540
79	5	1949	241723	5599	2.42191	1891.6	214.1	306
79	6	1128	241290	3502	2.38858	1905.1	216.6	91
79	7	751	241017	2516	2.35579	1933.2	218.9	11
79	8	672	240812	2277	2.32811	1946.5	221.1	15
79	9	753	240753	2570	2.33629	1960.1	223.4	57
79	10	1387	241010	4755	2.52487	1981.2	225.4	244
79	11	2775	241437	8919	2.80552	2005.5	227.5	523
79	12	4028	241828	12183	2.75624	2028.3	229.9	786
80	1	4965	242023	14797	2.77951	2046.5	233.2	1034
80	2	5822	242236	17189	2.76953	2055.6	236.4	1181
80	3	5126	242239	16605	3.17450	2069.6	239.8	1059
80	4	.	.	11469	.	.	242.5	.

TABLE A-4

EAST OHIO GAS COMPANY:
DATA ON RESIDENTIAL NATURAL GAS
CONSUMPTION AND ITS DETERMINANTS

Year	Month	Gas (mmcf)	Number of Customers	Revenue (10 ³)	Marginal Price	National Income (\$10 ⁹)	Consumer Price Index	Degree Days (°F)
70	1	33498	883448	27223	0.77107	778.8	113.3	1296
70	2	30823	884069	25262	0.79244	781.5	113.9	1239
70	3	27381	883890	22789	0.78829	787.6	114.5	1006
70	4	22036	882286	18841	0.80682	806.0	115.2	711
70	5	.	.	9701	0.80342	799.7	115.7	.
70	6	6313	876865	6204	0.80770	798.2	116.3	97
70	7	4754	874930	5071	0.83908	803.3	116.7	24
70	8	4108	873367	5164	0.85975	806.4	116.9	11
70	9	4181	873855	5309	0.87258	811.9	117.5	49
70	10	6542	878093	7355	0.88235	813.6	118.1	209
70	11	11761	882495	11968	0.89127	815.7	118.5	514
70	12	22397	886905	21466	0.89471	820.9	119.1	853
71	1	30528	888718	28832	0.92022	830.0	119.2	1177
71	2	33777	889504	32212	0.89821	833.2	119.4	1188
71	3	26854	889322	25932	0.92196	839.7	119.8	1032
71	4	21942	887607	21684	0.91456	844.4	120.2	841
71	5	13871	884821	14255	0.91707	850.0	120.8	464
71	6	7277	882113	8288	0.91888	870.1	121.5	147
71	7	4407	879209	5673	0.91724	859.2	121.8	13
71	8	4518	878291	5721	0.92105	867.6	122.2	11
71	9	3942	878901	5295	0.92349	871.5	122.4	38
71	10	6697	882588	7836	0.92111	874.8	122.6	116
71	11	10595	887766	11289	0.92705	879.4	122.6	436
71	12	23037	892955	22942	0.93603	890.4	123.1	766
72	1	26571	895156	26412	0.95726	898.9	123.2	994
72	2	30861	896486	30878	0.91111	908.5	123.8	1147
72	3	27882	896400	27577	0.96049	913.6	124.0	1032
72	4	22028	894923	22528	0.95917	919.4	124.3	747
72	5	11842	892417	12864	0.94894	924.0	124.7	380
72	6	6996	889640	8276	0.95469	922.9	125.0	160
72	7	5359	887287	6852	0.95268	932.9	125.5	78
72	8	4824	886161	6240	0.95103	940.0	125.7	30
72	9	4735	886962	6142	0.95907	946.8	126.2	61
72	10	9106	888563	10463	0.97392	967.0	126.6	290
72	11	16295	896281	17560	0.98582	977.6	126.9	619
72	12	26122	900099	27341	0.98339	983.6	127.3	845
73	1	27657	902268	28528	0.99547	989.1	127.7	1002
73	2	27052	903707	28235	0.97348	997.4	128.6	1050
73	3	22785	903681	24037	0.97941	1003.3	129.8	801
73	4	17386	901652	18729	0.98944	1011.6	130.7	510
73	5	12443	899512	13916	0.97703	1018.7	131.5	352
73	6	7634	897336	9132	0.97948	1026.6	132.4	129
73	7	4583	894700	6231	0.97426	1035.6	132.7	3
73	8	4115	893132	5631	0.96675	1047.3	135.1	6
73	9	4693	893897	6096	0.96461	1058.5	135.5	41
73	10	6579	895588	7958	0.96337	1090.8	136.6	154
73	11	13624	901436	14751	0.97997	1100.0	137.6	420
73	12	22511	905375	23587	0.53440	1107.1	138.5	776
74	1	28999	907723	22798	0.80160	1107.0	139.7	981
74	2	26689	908315	28141	1.09131	1113.4	141.5	1025
74	3	23928	908237	26019	1.01050	1117.1	143.1	906
74	4	19210	906435	22014	1.00686	1125.2	144.0	598
74	5	11728	903781	15397	1.07839	1135.2	145.6	350
74	6	6472	901314	9243	1.06956	1143.5	147.1	165
74	7	4535	899688	7032	1.07046	1159.5	148.3	26
74	8	4113	898687	6694	1.08238	1167.2	150.2	4
74	9	4892	899995	7779	1.13052	1178.0	151.9	91
74	10	10165	905626	14249	1.16515	1185.0	153.2	300
74	11	13204	909685	18057	1.42479	1184.5	154.3	542
74	12	24921	913248	35949	0.90517	1191.0	155.4	843

TABLE A-4
(continued)

Year	Month	Gas (mmcf)	Number of Customers	Revenue (\$10 ³)	Marginal Price	National Income (\$10 ⁹)	Consumer Price Index	Degree Days (°F)
75	1	27553	914619	33644	1.35776	1191.1	156.1	1024
75	2	26895	915590	36303	1.19101	1193.4	157.2	992
75	3	25797	916315	35401	1.26314	1195.7	157.8	948
75	4	22194	915057	30657	1.25590	1203.1	158.6	813
75	5	11392	912059	17349	1.22733	1214.3	159.3	423
75	6	5447	908777	9272	1.20828	1244.1	160.6	96
75	7	4286	906047	7681	1.19258	1238.9	162.3	22
75	8	3948	904283	7108	1.18308	1255.9	162.8	5
75	9	4841	905301	8252	1.23034	1270.9	163.3	96
75	10	8383	908935	13313	1.27745	1290.8	164.6	266
75	11	11150	913183	17277	1.52938	1300.2	165.6	439
75	12	24461	917819	38260	1.23692	1308.2	166.3	774
76	1	29034	920456	42811	1.57296	1320.8	166.7	1176
76	2	28837	921639	44143	1.40593	1331.4	167.1	1086
76	3	19175	920657	30592	1.42722	1341.9	167.5	725
76	4	16008	918436	25891	1.38885	1352.5	168.2	554
76	5	11562	916060	18856	1.36888	1362.9	169.2	401
76	6	6543	913000	11628	1.36263	1370.4	170.1	167
76	7	4361	910116	8676	1.36163	1380.8	171.1	14
76	8	4032	908314	8273	1.36837	1385.5	171.9	17
76	9	4752	908585	9409	1.40373	1391.7	172.6	97
76	10	9086	911904	17572	1.74489	1414.2	173.3	349
76	11	18673	915729	35360	1.90562	1432.1	173.8	741
76	12	28292	917739	55683	1.74129	1459.2	174.3	1125
77	1	35653	918607	69167	2.04696	1454.3	175.3	1398
77	2	32222	918481	64903	1.91160	1477.0	177.1	1311
77	3	20566	917410	43851	1.89231	1499.1	178.2	915
77	4	14476	915272	31188	1.86163	1510.1	179.6	568
77	5	8920	910813	20113	1.83397	1517.3	180.6	304
77	6	5662	905007	13674	1.76885	1524.3	181.8	141
77	7	4388	901031	10129	1.82240	1539.2	182.6	60
77	8	4644	900696	12471	1.80859	1540.7	183.3	15
77	9	4205	900845	10787	1.81768	1556.9	184.0	43
77	10	7180	902917	16488	1.85906	1577.0	184.5	219
77	11	10189	904830	22616	2.49461	1592.7	185.4	435
77	12	23420	909189	57313	1.24730	1609.2	186.1	848
78	1	30718	911450	56402	1.87095	1615.5	187.2	1245
78	2	31673	912160	65903	1.98929	1625.0	188.4	1365
78	3	26464	911487	55939	1.91824	1646.3	189.8	1174
78	4	17142	908261	37454	1.89750	1669.4	191.5	770
78	5	11868	905234	26835	1.84552	1682.1	193.3	376
78	6	5663	901693	14013	1.78001	1695.7	195.3	131
78	7	4298	898137	10409	1.76897	1730.0	196.7	25
78	8	3862	895858	10175	1.78140	1741.3	197.8	5
78	9	3658	894576	10224	1.82411	1756.1	199.3	23
78	10	7095	899243	17022	1.92435	1781.0	200.9	203
78	11	12030	905410	27436	2.62971	1801.4	202.0	491
78	12	21378	910752	56701	1.34724	1826.8	202.9	793
79	1	29799	913992	62800	2.62986	1834.3	204.7	1147
79	2	31085	915094	73287	2.23556	1851.4	207.1	1305
79	3	23147	914220	56144	2.22032	1872.1	209.1	981
79	4	16789	911563	43338	2.15392	1880.7	211.5	616
79	5	10264	912389	27544	2.08864	1891.6	214.1	421
79	6	6225	904709	17759	2.06390	1905.1	216.6	175
79	7	4182	901702	13476	2.05436	1933.2	218.9	40
79	8	4567	897481	14320	2.04459	1946.5	221.1	16
79	9	3556	900801	12097	2.16502	1960.1	223.4	49
79	10	7557	906674	22739	2.76281	1981.2	225.4	245
79	11	13777	913363	45804	3.37236	2005.5	227.5	537
79	12	19747	918222	71921	2.32729	2028.3	229.9	819
80	1	24612	921701	74151	3.49093	2046.5	233.2	1093
80	2	27723	923145	91427	2.71666	2055.6	236.4	1231
80	3	23638	922689	78226	.	2069.6	239.8	1106

degree days from the billing period. Then in month B a meter reading is taken that covers consumption in both months A and B. Any error made in the estimate of consumption in month A is eliminated in the month B billing. Month B's bill is calculated as the meter reading minus month A's estimate. Of course, meter readers work every month, so that half of the residential customers have meters read in month A and the other half in month B. The error or bias caused by this billing procedure on the monthly data of table A.2 is judged to be negligible. The primary reason is that estimates are made for hundreds of thousands of customers so that errors tend to cancel out between months A and B.

The third consideration of the billing data is that many customers, as many as a third, are on a budget plan so that equal payments are made every month in a year. This tends to make consumer response to price increases sluggish, suggesting that the income and price variables be lagged in modeling natural gas consumption. Preliminary analysis of the case studies suggested that this would not provide substantial improvements in models, so these data are not lagged.

Revenue data were obtained directly from the distribution utilities. These data are the sum of all billings for a particular month. Marginal prices were estimated using the AEP model of section 4.2.1 of this paper.

The national personal income data were obtained from Survey of Current Business, U.S. Department of Commerce. As mentioned in Section 4, monthly national income data are used, since state and city-level income data were not available on a monthly basis. The simple correlation between national income data and state of Ohio income data, compared on an annual basis, is 0.99 so that the national income data provide valuable information. The Consumer Price Index data were obtained from CPI Detailed Report, U.S. Department of Labor.

Finally, heating degree day data were obtained from Climatological Data, National Climatic Center, Asheville, North Carolina. Data from airports in or near Cincinnati, Columbus, Dayton, and Cleveland are included for the four distribution utilities.