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# SOUTHWEST ECONOMETRICS, INC.

#### UNDERSTANDING TRENDS IN TEXAS PER CAPITA WATER CONSUMPTION

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# 1996

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Prepared for

Texas Water Development Board

by

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#### I. INTRODUCTION

#### Recent Trends in Per Capita Water Consumption

Water planning entities in Texas and other areas of the Southwest have long held a common expectation concerning the water use rates of urban communities, namely that per capita water consumption would continue its long established upward trend. Such expectations were reinforced year after year as the data became available on current water use from municipalities which record and report their water use to planning agencies. This experience of the data consistently reinforcing the expectation continued unabated until the mid to late 1970s when use rates stopped rising. During the 1980s per capita water use began to decline and seems to have established a long term reversal of the upward trend.

The data show that municipal per capita water consumption in Texas "increased from about 100 gallons per capita per day (gpcd) in the post World War II era to levels slightly above 182 gpcd by the mid-1970s. Subsequent to then, average per capita use in the State had leveled out and...in 1978 averaged about 178 gpcd. By 1987,...consumption had fallen to about 170 gpcd, exhibiting a general declining trend over the ten-year period..." (see **Figure I-1**).<sup>1</sup>

These downward trends can be seen graphically in the data for several cities in Texas including Austin, San Antonio, Corpus Christi, Beaumont, Arlington and Pasadena (**Figure I-2**). These downward trends have major implications for water planners, especially since the planning horizons are very long in the discipline, reaching out some 40 years into the future in order to allow time for facility construction that often requires years of planning, permit processing, land acquisitions and construction.

Due to the importance of this long term trend to water policy and planning agencies, it is very essential to know the factors which are driving the downward trend in consumption rates.

<sup>&</sup>lt;sup>1</sup>Water for Texas: Today and Tomorrow - 1990, published and distributed by the Texas Water Development Board, Austin, Texas, December, 1990, p. 2-9.

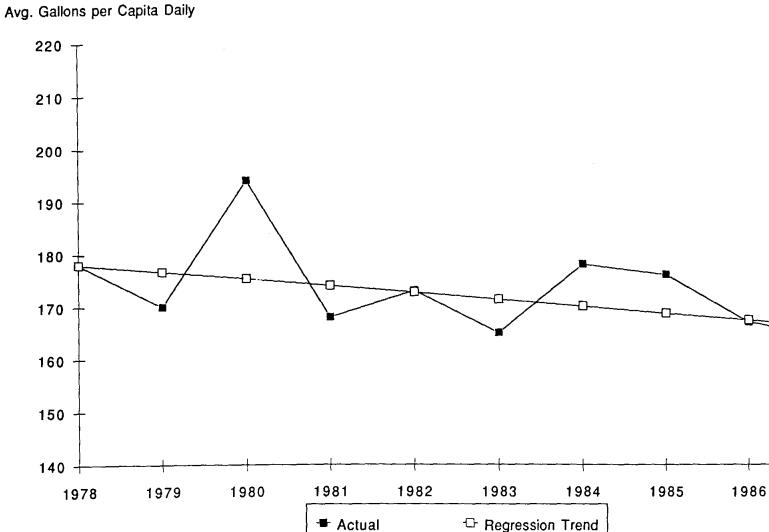


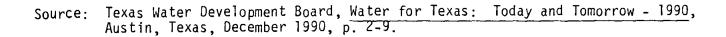
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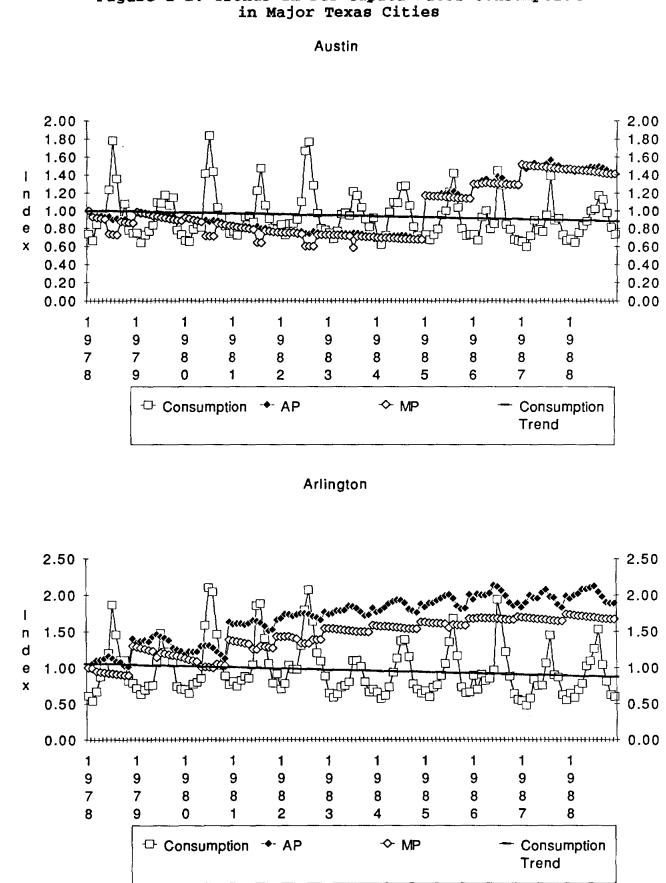
Texas Per Capita Municipal Water Use Trends





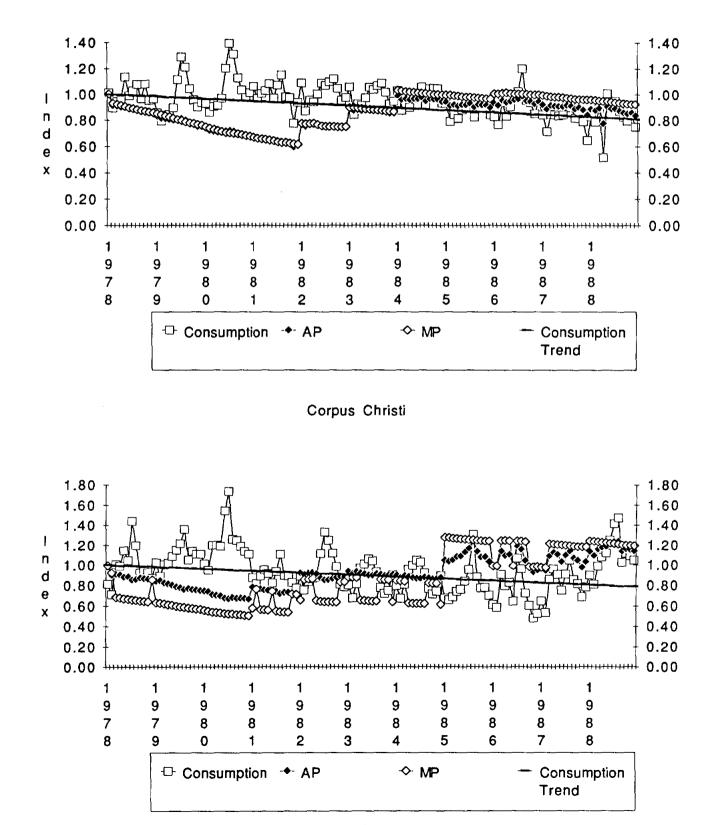


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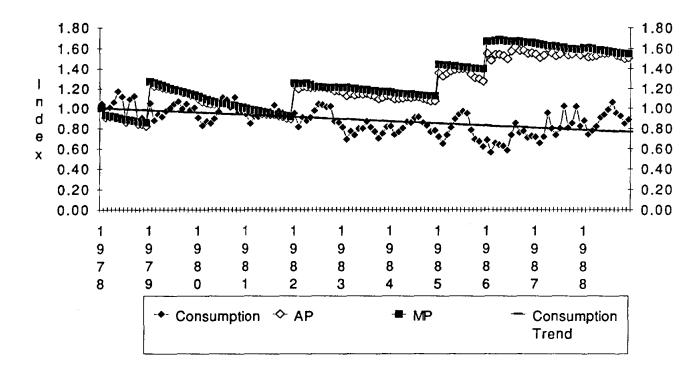


## Figure I-2. Trends in Per Capita Water Consumption

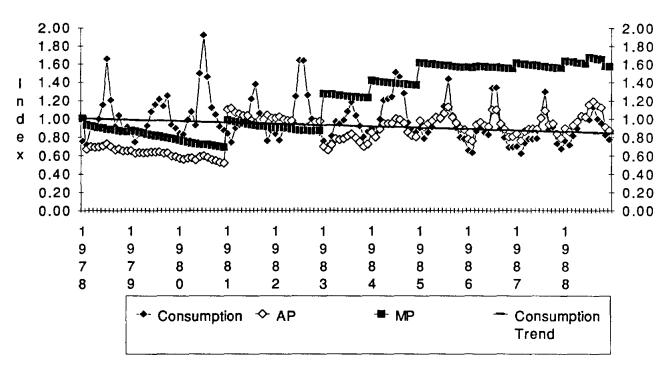








San Antonio



#### Price Structure Changes, Public Policy and Conservation

Several major events and public policy changes have, no doubt, had an important influence on water consumption in Texas including (1) the cyclical growth pattern of the Texas economy, (2) public policy in water and wastewater resources and (3) the development of a conservation ethic. During the 1970s and 1980s, Texas (and the nation) suffered three oil price shocks that brought long term shifts in the price structure of the economy which, in turn, have led to changes in energy costs of all types, changes in the use of materials and changes in the size of housing. These oil price shocks first increased the incomes of Texans relative to that of the nation in the 1970s and drastically reduced such in the 1980s following the oil price collapse in 1986.

The rapid growth of Texas communities in the late 1970s and early 1980s caused most communities to overbuild water and wastewater facilities since they expected growth to continue. Such overbuilding led to cost and rate increases above the long term trends.

Another important change that occurred during the period of the late 1970s and early 1980s was a major shift in Federal wastewater policy, namely the drastic reduction in funding for wastewater treatment plants, plants which were mandated to be built in order to meet Federal clean water standards. This reduction in Federal funding that had for years, stayed at 90% of the cost of wastewater treatment plants shifted the cost burden to state and local agencies. The end result was a significant increase in rates for wastewater services in many Texas communities. Federal Clean Water and Safe Drinking Water Acts also imposed higher costs on utilities by increasing the standards for the provision of wastewater and water services.

Another major public policy change occurred at the Texas State government level. A new statute amending the Texas Water Code was adopted by the Legislature in 1985 which, among other things, established a new set of financing mechanisms to provide low cost financing of local water projects by the extension of the state's credit capacity and cost of money to local government agencies. As a requirement of obtaining such State assistance, local entities were required to develop and submit a conservation plan that needs to include certain characteristics established by rule of the Water Development Board and/or by State statute.

Another factor, which is difficult to quantify, but which may indeed explain some of the trends at work, is the development of a broad based "conservation" ethic concerning the use of natural resources. The series of oil crises,

water shortages and related general environmental awareness has no doubt been responsible for the development of this conservation ethic.

#### Purpose of This Study

The purpose of this study is to determine the several factors that underlie and explain the recent downward trends in per capita municipal water consumption in Texas. Further, the purpose of the study is to quantify the relationship between these various factors and per capita water use.

#### II. REVIEW OF LITERATURE ON WATER CONSUMPTION

#### Econometric Models of Consumption

Econometric methods have been used extensively over the last thirty or so years to estimate the relationship among various economic, climatic and sociological factors and water consumption. The literature of such studies is dominated by the work of economists and engineers.

The economics profession naturally thinks of the type of problem posed by this study in terms of micro economic theory, namely the supply and demand for a consumer item. Therefore, a review of the literature in economics publications will usually turn up a list of studies that concern attempts to estimate "demand functions" that relate water consumption to price, income, structure of the decision making unit (households), weather and climatic factors, and perhaps the structure of price.

A literature review of the topic of explanations of water consumption will also turn up numerous studies completed by engineers. Because engineers are less concerned with theoretical underpinnings than are economists, one often finds in these studies a process modeling approach that is more rigorous as a perceptive tool than as a descriptive tool. That is, the only ability to test the explanatory power of the model against historical experience in such models is by comparative statistics and visual inspection. Such processing models have their greater strength in organizing the informed judgment of the authors into a system that allows the inclusion of new influences not in the historical data. For example, one can model the expected influences of conservation programs on future water consumption by use of a process model even though conservation programs did not exist during the historical period from which economists derive data for econometric models. More discussion of this topic is included in the next section of the report.

The review of literature here focuses primarily on econometric models of water consumption since it is our main purpose to identify and quantify the factors that explain the recent trends in per capita water consumption. Because of the potential importance of conservation programs on the long term future of water consumption, however, some attention is also given to processing models and methods for considering influences in the future that do not exist in the historical experience.

A comprehensive review of the econometric water demand models was completed by the U.S. Army Corps of Engineers in

1984.<sup>2</sup> This publication not only reviews the work to date but focuses on the question of price elasticities of demand in water consumption. Several recent studies have been completed that focus directly on Texas.<sup>3</sup>

Boland et al. reviewed more than 50 substantial studies of the response of municipal and industrial water use to price. The review included mostly work done on the topic since the 1960s. Not only did Boland et al. review the content of the studies but made judgements about the statistical rigor of the studies and drew conclusions about the range of price elasticities that characterize demand by summer and winter use by region of the U.S. by user class. The studies typically included explanatory variables of price, number of households, persons per household, household income, property value, irrigable area and climate.

Boland et al. found that there had been a number of conventions used in the specification of the consumption variable and the price variable, as well as others. Important in explaining the overall variation in consumption in residential use is the number of households, and while often a statistically significant variable, price makes a relatively small contribution to the overall explanation of the variance in consumption.

Results of the Boland et al. study indicate that price elasticities of demand for water are likely to be in the following ranges:

<sup>2</sup>Boland, John, Bondedykt Dziegielewski, Duane Baumann and Eva Opitz, Planning and Management Consultants, Ltd., Influence of Price and Rate Structures on Municipal and Industrial Water Use, for U.S. Army Corps of Engineers, Institute for Water Resources, Carbondale, IL, June 1984. <sup>3</sup>The most comprehensive studies done in Texas in recent years include two state-wide studies by Ronald Griffin at Texas A&M and a focus study of the Texas Mexican border water demand by Milton Holloway. The Griffin studies are: Griffin, Ronald C., and Chan Chang, Community Water Demand in Texas, Texas Water Resources Institute, Texas A&M University, April 1989, and "Community Water Demand: New Specifications," Western Agricultural Economics Association, Honolulu, Hawaii, July 10-12, 1988. The Holloway study is: Holloway, Milton L. and Doug Tharp, "A Methodology for Determining Ability to Pay: For Use in the Implementation of the Economically Distressed Areas Water Assistance Program, " for the Texas Water Development Board, Austin, Texas, March 1990.

	Elasticity				
	Long Run	<u>Short Run</u>			
Residential (winter) Residential (summer)	0.0 to -0.10	NA			
Eastern U.S. Residential (sprinkling)	-0.50 to -0.60	NA			
Eastern U.S.	-1.30 to -1.60	NA			
Western U.S.	-0.70 to -0.90	NA			
Residential Average	-0.20 to -0.40	0.0 to -0.30			
Commercial	-0.20 to -1.40	NA			

Griffin and Chang, in the 1989 study, estimated demand functions for municipal water use in Texas communities. The data base included 221 communities with data for the period 1981-1985.<sup>4</sup> Griffin found that average price is empirically preferred to marginal price, and that the monthly price elasticities are on the order of -.14 in the winter and -0.28 to -0.37 in the summer, measured at the means of monthly consumption. Griffin tested prices in both real and nominal terms, but did not express a strong preference as to which to use. He also included a sewer price in the definition of water price and found that it should be included. That is, he performed a test of sorts and concluded that the data suggest that consumers don't know water and wastewater prices separately, or individually for that matter, but instead are aware of only the monthly bill. The monthly bill usually includes both water and wastewater.

The Holloway and Tharp study of 1990 had the purpose of estimating the ability of communities in the Texas/Mexico border area to pay for water and wastewater services based on what persons of similar economic circumstances were in fact paying. The modeling involved the estimation of regional demand functions from cross-sectional data derived mostly from the Bureau of the Census 1980 public use sample data. Price elasticity estimates are in the neighborhood of -0.80 for the communities included in this study.

#### Measuring Conservation Effects

There is naturally a strong interest of policy makers and planners in knowing the extent to which conservation programs of one type or another have any impact on water consumption. It seems clear, for example, that an intensive public awareness program during a drought period, or perhaps the initiation of an odd-even day watering scheme, has a

<sup>&</sup>lt;sup>4</sup>Several major MSA cities were eliminated from the data set, however, including San Angelo, Plano, Pasadena, Mesquite, McAllen, Lubbock, Houston, Grand Prairie, Dallas, Ft. Worth, Baytown and Abilene.

significant impact on water use. It is much less clear whether such programs of awareness or regulatory restriction have any lasting impact beyond the current crisis. On the other hand, incentive programs for the installation of lowflow shower heads and toilet dams are sure to have a measurable, long term impact on non-sprinkling water use.

A statistically valid method of evaluation of the contribution of conservation programs would be to complete a controlled experiment where selected households would be identified for record keeping over a long period of time. The group would be offered the benefits of a particular conservation program, e.g. free low-flow shower heads and toilet blocks. Those who took advantage of the program would be put in one class and those who did not, in another. Records on consumption, price, income, household size, weather and climate would be maintained over a long enough period of time to determine the behavior of the group who elected to take advantage of the program. One would want to know, for example, whether the shower heads were replaced with regular models and whether the units stayed in place after the house was sold, etc. That is, a well designed test of the difference made by the conservation program would require some control over the data in order to apply normal statistical methods to the question. There are no such studies in the literature that we are aware of.

Another method of analysis would be applied at the community level rather than the individual consuming level. That is, it is possible to statistically compare the water use of communities over time that have, among all the variables that tend to explain consumption, a set of communities that have conservation programs with those who do not. There is no evidence in the literature that such a study has been completed.

#### III. METHODOLOGY FOR EXPLAINING PER CAPITA WATER CONSUMPTION TRENDS

The method of analysis for explaining the Texas trends in per capita water use selected for this study is that of multiple regression. Time series and cross-sectional data were combined in a data base of per capita consumption for 72 communities for the time period of 1978-1988. The communities were grouped into various regional groupings and equations were estimated for each. The equations allow one to explain recent trends in consumption and to use these equations for forecasting future consumption. The data base and model specification are explained, respectively, in Chapters IV and V.

#### Definitions

There have been a number of conventions developed from studies of water consumption that have been considered in developing the current study. Some have been accepted and others have not. The issues surrounding the selection of variables, geographical coverage and time period of analysis are discussed in this section. Specific definitions of variables are in Chapter IV.

#### Consumption

Three types of consumption data are typically found in the literature on water demand. "Metered data" are most often used since they are unique to the consumer decision making unit (household or business) and are readily available from utility records. A second type is that of survey results where consumption is a derived calculation based on "reported expenditures" by the survey respondent. A third type is a "calculated disappearance" quantity that may be derived from gross withdrawal data reported by a utility serving a community.

The metered data has the obvious strength of being directly derived from the behavior of the decision making unit, the consumer, who makes the choices of budgeting and purchasing that is of the greatest interest in demand analyses. One weakness of this data is that billings information is collected by address, not household or business, so that the behavior of the decision maker over a period of time is not preserved in the data. Second, if one is interested in the explanation of, and forecast of, aggregate water consumption for a city, county or state, individual billing data are massive amounts of information to manage. Still another, and perhaps the most important weakness of billings data, is the absence of associated income, household size, housing characteristics or other likely independent variables for use in econometric analyses.

Survey results that allow derivation of consumption via billings and rate structure information have the strengths of going directly to the decision making unit to gather the data on consumption, and at the same time gathering income, household size, housing characteristics, etc. that are needed for econometric analysis. One weakness of such data is that respondents do not usually know quantity consumed and may be able to provide only "ball park" expenditure information from which to derive consumption. A second weakness is that of cost. If one is interested in explaining and forecasting community and state level water consumption, survey data are expensive to obtain and often impractical to gather.

Gross withdrawal derived data have the strength of accounting for the total water consumption of a community. One of the weaknesses is that it is impossible to capture the direct association of the individual decision maker's water consumption with income, household size, housing characteristics, etc. that are needed to distinguish water consumption behavior in the context of budgeting and consuming decisions. Therefore, this type of data only allow analyses <u>among</u> communities where each community is, in essence, treated as a decision making unit. This allows the consumption of water to be associated with income, household size, climate and weather factors and price at the community level.

The consumption data selected for this study are a mixture of two of the three types discussed above. The Water Development Board has for years, maintained a community level data base of water consumption consisting of derived annual per capita consumption and a monthly distribution function that allows the derivation of per capita consumption by month by community. This data base is constructed by first calculating the total disappearance of water within a community based on the net of gross withdrawals and wholesale sales of the utility that serves the community. Since large industrial users are usually independent of the utility serving the community, the resulting data are residential plus small and medium commercial users divided by the number of people in the community. While this characterization of "per capita water consumption" has some obvious weakness, such as the variance due to the number and character of commercial consumers in the data, from an overall perspective, these data are the best available for this study. First, the data base represents the combined experience of several professionals over a long period of time with knowledge of each community in the set. Second, the data are based on utility reported information that captures accurately the total water

consumption by month for each community. For purposes of explaining and forecasting community level consumption for the State of Texas, this is the preferred data base.

#### Price

The problem of how to characterize the price of water for demand analyses is not an easy one. First of all, consumers usually are faced with a combined bill for water and wastewater so that price for the individual service of providing water and wastewater on a monthly basis is not recognized by the typical consumer. Second, there is the problem of whether one is interested in the average or the marginal price which in today's utility pricing are often guite different. Another problem is a conceptual one for demand analysis purposes. Theoretically one expects to use the marginal price for demand analysis because the micro economic theory of markets tells us that prices are determined at the margin or, said another way, that individual consumers are always faced by the marginal price when deciding whether or how much additional service or commodity to purchase.

There are many practical problems that cloud this issue. It is practically difficult to obtain either average or marginal prices for individual consumers since the only access to such information is through surveys or individual billings data. Community averages can be derived from total consumption and total utility revenues, but the marginal price that corresponds to that average must incorporate the rate structure of the utility.

The data chosen for the study are described in detail in Chapter IV, but as a general matter it was determined that we should test both the average and marginal price and that such could be derived by combining the average monthly per capita consumption data with average number of persons per connection and the rate structure for the utility, for each year of the period 1978-1988. That is, average and marginal prices at the average per capita consumption level by month by community can be derived for each community in the data set.

#### Income

A significant explanatory variable for explaining the level of water consumption among communities and over time is income. This variable captures a combined set of factors that relate to the housing and commercial building stock in a community that, in turn, has much to do with water consumption. The idea is that income of a community determines the size and character of housing for the residential sector and building space for the commercial sector. Implicit in the purchase of such building space is the number of bathrooms and showers, as well as the size of lawn which requires irrigation. As income rises over time, water use and water using capacity also tend to rise, other things equal.

The only comprehensively available source for income data is that of the Bureau of Economic Analysis, U.S. Department of Commerce. This data is available at the county level for years up through 1988. This data was selected to represent each community within each county of the data set.

#### Weather

Any study seeking to explain variations in water consumption will need to take account of weather conditions that have an obvious influence on short term variations in consumption. Weather conditions may also influence consumption over the long term within the cycles of weather patterns that sometimes last for years.

Fortunately, Texas has a large number of weather stations located throughout the State, such that there is a data gathering system near almost every city within the Metropolitan Statistical Areas (MSA) of Texas. These data are public information and available through organized data systems such as the Texas Natural Resources Information System (TNRIS).

The variables of interest from weather station sources include temperature and precipitation. The expectation is that summer sprinkling water use, in particular, is heavily influenced by the extent of hot, dry days when transpiration rates are high. While forecasting by use of equations estimated from historical data will normally assume normal long term weather conditions, it is essential that adequate weather representations be included in the use of econometric models of historical consumption. Typical representation of weather includes maximum or average daily temperature and number of dry days. The specific form of these variables for the current study are discussed in Chapter IV.

#### Conservation

The term "conservation" has a variety of popular uses, but there is no commonly accepted definition of the term for analytical work. One confusing area concerns the use of the rate structure for "conservation" purposes. If asked to list and describe the conservation programs being implemented, utility employees often list a change in rate structure that has been revised from a declining or flat price per unit to an increasing block structure. One might well classify the use of an increasing block rate structure as a conservation program if the increasing structure, in fact, bears no correspondence with marginal costs. The matter is complicated by conventions of pricing by regulated utilities.

Other common responses to questions about the definition of conservation programs include education, rationing during drought conditions, and subsidy programs to encourage investment in water savings technologies such as low-flow shower heads and toilet dams.

For our purposes in this study we have included all rate structures in the calculation of average and marginal prices for each city included in the analysis, regardless of whether the utility listed the rate structure as a conservation program. Education programs and subsidy programs to encourage installation of water conserving technology have been classified as conservation programs. Rationing during drought periods has been eliminated from the analysis altogether.

Conservation, as defined above, has been included in the analysis by testing whether there is a statistically discernable difference in per capita consumption due to the presence of a program. That is, a conservation variable has been defined and included in the regression analysis. The method of inclusion was to identify which communities have conservation programs and at what point in time they were begun.

#### Extent of Geographical Coverage

This study of per capita water use trends distinguishes municipal water use from other types of use, namely, industrial, electric utility and agricultural uses. The analysis deals only with municipal water use which includes the retail water sales of utilities to residential and small commercial classes of users. Large commercial/industrial users typically provide their own water and wastewater service or purchase from a utility on an individual contract basis.

Since the study is designed to deal with municipal water use only, a decision was made to limit the data gathering and analysis to the set of 72 cities that make up the 28 MSAs in Texas. These cities account for about 85% of the municipal water use in Texas.

#### Importance of Diversity Among Cities

One of the strengths of econometric analyses is that one can be definite about the population to which the analysis is applicable. That is, the statistical tests that allow one to have confidence in the value of parameters estimated statistically apply only to the population from which the data were drawn. In this case, the results will apply to each and every city included in the analysis, but not to others. Therefore, the results will be strictly applicable to all the utilities within the MSAs of Texas that consume 85% of the municipal water in Texas; no more or no less.

The inclusion of each city in all of the MSAs in Texas insures that the diversity of climate, geology, culture and costs get considered. It insures that we will be able to derive meaningful information for the full range of diversity among Texas cities.

#### Changes Over Time

Most dynamic processes involve some time lapse before all of the influence of a prior change is fully played out. For example, if relative prices change, influencing the economics of choice between two consumer goods, the ability to take advantage of the favorable price change may involve some investment in new equipment, such that the change in consumption patterns is not really evident until later periods. For this and related reasons it is usually advisable to include several time periods in an analysis.

This study of per capita water use trends makes use of 11 years of data (1978-1988). One year completes a seasonal cycle of water use patterns that are influenced by weather and plant growing seasons. The inclusion of 11 years of time series data allows the analysis to span a major rise and fall of the Texas economy and to consider the lag effects that may accompany consumer response to price and income changes of the late 1970s and early 1980s.

#### Strengths and Weaknesses of Econometric Analyses

The strengths of econometric analyses fall into two categories. The first has to do with the degree of confidence one may have in the parameters that quantify relationships among variables. For example, the measure of consumer response to price changes can be estimated using such analyses, and these price effects may be separated from another influence such as income changes. Not only can one separate the effects of two such influences, but he can also derive statistical tests that allow a measure of confidence in the estimate.

Another strength of econometrics is that it allows one to analyze an enormous amount of data efficiently. The current study, for example, involves 72 cities and 132 monthly observations each.

The only significant weakness of econometric analysis for the purposes of this project is that it is limited to factors and relationships that are present in the historical period. One cannot measure the future effects of a conservation program being put in place today using econometrics if there is no comparable set of programs included in the available historical data. Other modeling approaches will be required for such problems. For example, one could construct a prescriptive model based on cost minimizing behavioral assumptions for such a question.

#### Formulation of an Econometric Model of Texas Water Consumption

There are two sources for developing a hypothesized mathematical model of Texas municipal water consumption. One is from economic theory of consumer behavior that provides the information that consumers' economic choices tend to follow general rational responses, such as decreasing consumption when prices rise and increasing consumption when incomes rise (other things equal). From economic theory we bring the following information to the current problem:

- quantity consumed is inversely related to price (inflation removed);
- (2) quantity consumed is directly related to incomes (inflation removed);
- (3) quantity consumed is directly related to family size or persons per household;
- (4) quantity consumed is directly related to temperature and plant moisture stress; and
- (5) there are complementary and substitute consumer products that may come into play when relative prices change.

Another important source of information for formulation of the current model of water consumption is the literature. A number of "hints" and "leads" come from past efforts to solve similar problems. From the literature, for example, we have expectations about the range of price elasticities that may come from the current work (see Chapter II). We also have accounts of variable definitions, mathematical formulations and statistical test results obtained by others. Based on both sources of information discussed above, the following general model was formulated for the current analysis:

Q <sub>i</sub> = the per capita consumption in time period i (i = 1132), for MSA j (j = 128)	i
NP <sub>ij</sub> = number of persons per connection in time period i and MSA j	
<pre>AP<sub>ij</sub> = real average price per 1,000 gallons per month in time period i and MSA j</pre>	
<pre>MP<sub>ij</sub> = real marginal price per 1,000 gallons per month in time period i and MSA j</pre>	
I <sub>ij</sub> = real per capita income in time period i and MSA j	
$T_{ij}$ = temperature by month in time period i and MSA j	
DD <sub>ij</sub> = number of dry days per month in time period i and MSA j	

Our prior expectation is that the signs on coefficients estimated for these variables will be positive (+) for NP, negative (-) for AP and MP, positive (+) for I, positive (+) for T and positive (+) for DD.

An alternative specification of a model would include a representation of a supply function and the equation specification would be a simultaneous equation set. Such a specification is needed, conceptually, to separate shifts over time in a demand function from movements along a demand function that may accompany supply function shifts. This problem in applied economics is known as the identification problem. The literature on the topic suggests that attempts to estimate a simultaneous equation set is unlikely to succeed. As discussed later, an attempt to estimate a simultaneous equation set did not prove successful here either.

#### IV. DATA BASE

#### Metropolitan Statistical Areas

Four types of data constitute the data base for evaluating trends in per capita municipal water use in Texas: water consumption, water price, income and climate. The Metropolitan Statistical Area (MSA) was the basic demographic unit by which these data were collected. The time period of record is 1978 through 1988.

#### Geographic Coverage

Cities (utilities) for which data were collected represent the cities within each MSA required to total at least 80 percent of the population of each of the 28 MSAs in Texas. This selection process yielded 72 cities with populations within a wide range (**Table IV-1**). Monthly data for 1978-1988 exist for each city in the data base.

#### Water Consumption

Monthly water consumption data were derived from the data base of the Water Development Board, which includes annual average per capita consumption per day, and the monthly distribution of the annual average daily consumption. The Water Development Board's population estimates and the number of residential connections reported by each utility were used to convert per capita daily consumption to household consumption in gallons per month. Household consumption is used instead of per capita consumption in order to be consistent with billing practices and to measure economic responses at the basic decision making unit.

#### Income

Income data are derived from the county per capita annual income estimates of the Bureau of Economics Analysis (BEA), regional Economic Measurement Division, as updated in May 1990. The BEA's income data were deflated by the consumer price index for the South (1982-84 = 100), published by the Bureau of Labor Statistics.

#### Average and Marginal Price

Average and marginal prices were derived for each city in the analysis through municipal rate schedules on file

### TABLE IV-1. MSAs, CITIES AND COUNTIES

MSA#	MSA	CITY#	UTILITY	COUNTY	MSA#	MSA	CITY#	UTILITY	COUNTY
1	Abilene	2	Abilene	Taylor	12	Galveston	219	Friendswood	Galveston
2	Amarillo	14	Amarillo	Potter	12	Galveston	227	Galveston	Galveston
3	Austin	30	Austin	Travis	12	Galveston	350	League City	Galveston
4	Beaumont/PtArthur	43	Beaumont	Jefferson	12	Galveston	602	Texas City	Galveston
4	Beaumont/PtArthur	476	Port Arthur	Jefferson	13	Houston	42	Baytown	Harris
5	Brazoria	13	Alvin	Brazoria	13	Houston	285	Houston	Harris
5	Brazoria	18	Angleton	Brazoria	13	Houston	456	Pasadena	Harris
5	Brazoria	72	Brazoria	Brazoria	13	Houston	130	Conroe	Montgomery
5	Brazoria	118	Clute	Brazoria	14	Killeen/Temple	322	Killeen	Bell
5	Brazoria	217	Freeport	Brazoria	14	Killeen/Temple	597	Temple	Bell
5	Brazoria	338	Lake Jackson	Brazoria	14	Killeen/Temple	134		Coryell
5	Brazoria	457	Pearland	Brazoria	15	Laredo	347	Laredo	Webb
6	Brownsv/Harlingen	80	Brownsville	Cameron	16	Longview/Marshall	321	Kilgore	Gregg
6	Brownsv/Harlingen	265	Harlingen	Cameron	16	Longview/Marshall	367	Longview	Gregg
7	Bryan/College Station	82	Bryan	Brazos	16	Longview/Marshall	388	Marshall	Harrison
7	Bryan/College Station	124	College Station	Brazos	17	Lubbock	370	Lubbock	Lubbock
8	Corpus	135	Corpus	Nuecas	18	McAl/Edin/Mission	182	Edinburg	Hidalgo
9	Dallas	472	Plano	Collin	18	McAl/Edin/Mission	376	McAllen	Hidalgo
9	Dallas	98	Carrollton	Dallas	18	McAl/Edin/Mission	397	Mercedes	Hidalgo
9	Dallas	151	Dallas	Dallas	18	McAl/Edin/Mission	408	Mission	Hidalgo
9	Dallas	230	Garland	Dallas	18	McAl/Edin/Mission	463	Pharr	Hidalgo
9	Dallas	245	Grand Prairie	Dallas	18	McAl/Edin/Mission	638	Weslaco	Hidalgo
9	Dallas	298	lrving	Dallas	19	Midland	404	Midland	Midland
9	Dallas	401	Mesquite	Dallas	20	Odessa	438	Odessa	Ector
9	Dallas	498	Richardson	Dallas	21	San Angelo	529	San Angelo	Tom Green
9	Dallas	159	Denton	Denton	22	San Antonio	530	San Antonio	Bexar
10	El Paso	189	El Paso	El Paso	23	Sherman/Denison	158	Denison	Grayson
11	Fort Worth	115	Cleburne	Johnson	23	Sherman/Denison	556	Sherman	Grayson
11	Fort Worth	25	Arlington	Tarrant	24	Texarkana	429	New Boston	Bowie
11	Fort Worth	44	Bedford	Tarrant	24	Texarkana	601	Texarkana	Bowie
11	Fort Worth	193	Euless	Tarrant	24	Texarkana	628	Wake Village	Bowie
11	Fort Worth	213	Fort Worth	Tarrant	25	Tyler	613	-	Smith
11	Fort Worth	249	Grapevin <b>e</b>	Tarrant	26	Victoria	624	Victoria	Victoria
11	Fort Worth	261	Haltom City	Tarrant	27	Waco	47	Bellmead	McLennan
11	Fort Worth	293	Hurst	Tarrant	27	Waco	626	Waco	McLennan
11	Fort Worth	435	North Richland Hills		27	Waco		Woodway	McLennan
L					28	Wichita Falls	654	-	Wichita

residential and commercial use. Weighted marginal and average prices were derived by the relative mix of residential and commercial water connections reported annually to the Water Development Board by each utility.

#### Climate

National Weather Service (NWS) data regarding precipitation and temperature for selected Texas weather stations were acquired from TNRIS. Data were selected from the NWS station nearest the city for which data exist for the period of record 1978-1988.

Temperature data are the average monthly temperatures at the NWS station nearest the city, and for which data exist for the period of record 1978-1988. Average monthly temperature is the mean of the average daily high temperature and average daily low temperatures as reported in two separate data bases at TNRIS.

Precipitation data are the total number of days in a month with less than 0.25 inches of precipitation at the NWS station nearest the city, and for which data exist for 1978-1988.

#### V. THE MODEL

#### Important Factors Affecting Consumption

The model for evaluating trends in per capita municipal water use in Texas was specified as a demand model, or a model in which the effect of price on consumption is measured. Important factors other than price which affect consumption are income, number of persons per household, average monthly temperature, the number of days per month without significant rainfall and the level of commercial development. Not all of these factors affect each city or MSA uniformly but are always important.

#### Nine Regional Models

Early analysis revealed that water consumption is better evaluated on a regional basis. Therefore, equations were estimated for nine regions of Texas. MSAs were grouped together based on a combination of criteria: location with respect to vegetational and geological designations, general precipitation patterns based on data from 1950 through 1981, commercial distinctiveness and city size.

Twenty-eight MSAs were grouped into the nine regions (**Table V-1**). The Metroplex (Dallas-Fort Worth) area constitutes two separate regions, each of which includes cities from both the Dallas MSA and the Fort Worth MSA. The distinction is due to suburban location of cities in both MSAs.

#### Cross-Sectional/Time-Series Combination

The period of record is 1978-1988 from which a data base was constructed containing monthly information for each city within each region. This time series affords the ability to analyze the response over time of water consumption to the explanatory variables of the model.

Grouping several MSAs, each of which contains one or more cities, allows cross-sectional analysis by which to examine the relationship between consumption and the explanatory variables for multiple locations within one period of time. The combination of time-series and crosssectional data for analysis allows for explaining regionwide structural relationships and changes in those relationships over time.

REGION	<u>MSA #</u>	MSA
West	10	El Paso
	20	Odessa
	19	Midland
	17	Lubbock
	2	Amarillo
Rolling Plains	28	Wichita Falls
	1	Abilene
	21	San Angelo
	15	Laredo
$Metroplex^1$	9	Dallas
	11	Fort Worth
Metroplex Suburban <sup>2</sup>	9	Dallas
	11	Fort Worth
Central	23	Sherman-Denison
	27	Waco
	14	Killeen-Temple
	7	Bryan-College Station
I-35 South	3	Austin
	22	San Antonio
	26	Victoria
Southeast	5	Brazoria
	12	Galveston
	8	Corpus Christi
East	24	Texarkana
	25	Tyler
	16	Longview-Marshall
	4	Beaumont-Port Arthur
	13	Houston
Valley	6	Brownsville-
		Harlingen
	18	McAllen-Edinburg-
		Mission

TABLE V-1 MSA Groupings for Regional Models

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<sup>&</sup>lt;sup>1</sup>Includes cities of Fort Worth, Arlington, Dallas, Plano, Carrollton, Irving and Richardson.

<sup>&</sup>lt;sup>2</sup>Includes cities of Cleburne, Bedford, Euless, Grapevine, Haltom City, Hurst, North Richland Hills, Garland, Grand Prairie, Mesquite and Denton.

#### Functional Forms of Nine Regional Models

In general form, the model for water consumption is specified as:

CONS = f (MP, FAMLYINC, TEMP, DAYS, COMPROXY, DSEAS, Dn) where:

CONS = per household water consumption in gallons per month;

MP = weighted marginal price in dollars per thousand gallons;

FAMLYINC = per capita income multiplied by the number of persons per residential connection, in dollars;

TEMP = the average monthly temperature in degrees
 Fahrenheit;

DAYS = the number of days with precipitation of less than 0.25 inches;

COMPROXY = the fraction of total water connections attributable to commercial use;

- DSEAS = dummy variable which distinguishes summertime consumption from consumption in the rest of the year.
- Dn = locational dummy variables for MSA number n which distinguishes one MSA from another (number designations in Table V-1).

Three functional forms of each regional model were estimated econometrically using the Statistical Analysis System (SAS) regression procedure. The three forms are linear, log-linear and log-log, all results of which are in Appendix A.

Table V-2 contains the parameter estimates, t statistics, F test and number of observations (n) for the preferred functional forms for all nine regions. The loglinear form provided the best results for all regions except the East and the Rolling Plains. The Rolling Plains region was estimated in linear form. The East region was estimated in log-log form.

All parameter estimates in all nine equations are statistically significant with signs which are intuitively correct. The relationship between consumption and price is inverse, and the relationship between consumption and income is direct. The signs in the TEMP and DAYS parameters are all positive, indicating that higher monthly average temperatures and a larger number of days without significant rainfall tend to induce higher water consumption, all other variables remaining unchanged.

insignificant in early equations in which the sign on the parameter was negative.

#### Explanation of Historical Water Consumption

Data from seven cities from six separate regions were used to indicate the performance of the models in explaining historical consumption. **Figures V-1** through **V-7** show the actual vs. predicted values of dependent variables for water consumption in El Paso, Abilene, Dallas, San Antonio, Austin, Corpus Christi and Houston. Actual data are cityspecific. The models are the respective regional models contained in **Table V-2**.

Variations in consumption were explained well for El Paso (Figure V-1), Abilene (Figure V-2), Dallas (Figure V-3), San Antonio (Figure V-4), Austin (Figure V-5) and Houston (Figure V-7) with predicted values approaching the actual values even in the summer peak consumption periods. Consumption patterns in El Paso, Abilene, Dallas, San Antonio, Austin, Corpus Christi and Houston appear to typify the patterns of their respective models. That is, the combined effects of variation in price, income, dry days, temperature and concentration of commercial water users do a good job of explaining the variation in monthly consumption in these cities using the applicable regional model. Only Corpus Christi (Figure V-6) reflects results atypical of its regional model.

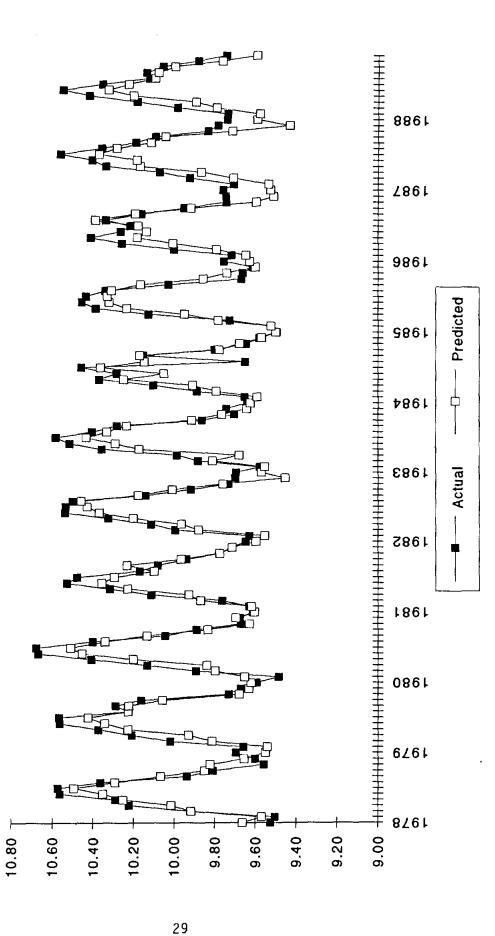
The predicted values in Corpus Christi vary most from actual values in 1980, 1986 and 1988. Directional patterns show underestimation in 1980 and 1988, and overestimation in 1986. **Figure V-6** shows a tendency to over-predict in the last half of the period, which could be the result of the model's failure to capture the effect of conservation programs in Corpus implemented in the last half of the 1980s. The strong reversal to a high level of underestimation in 1988, however, seems to discount this possibility.

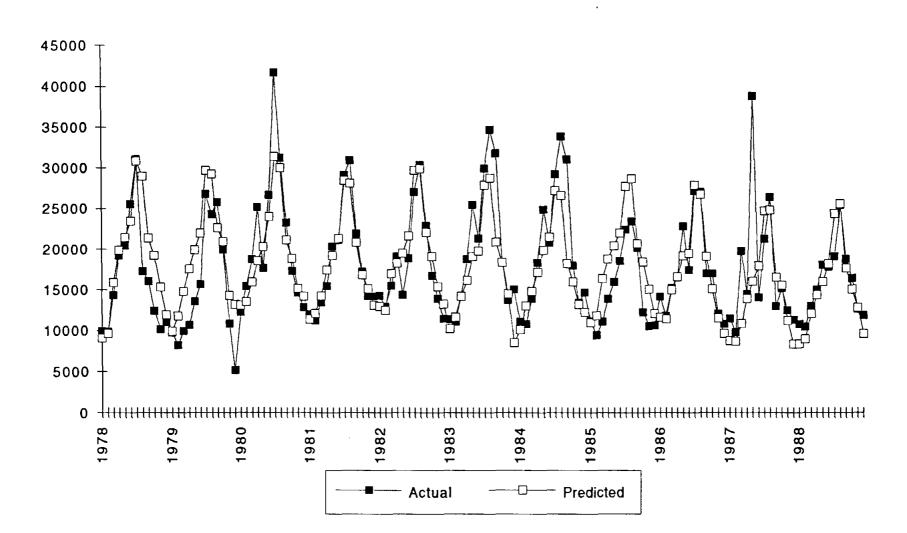
As mentioned earlier, attempts to estimate a conservation parameter did not prove successful. That is, the set of explanatory variables do not include conservation. Since a number of cities implemented conservation programs during the 1980s, we are interested in checking the patterns of predicted vs. actual consumption residuals to see if unexplained variation has a long-term trend that possibly could be explained by conservation programs.

An examination of residuals (**Appendix B**) indicates that the Southeast regional model explains variation in Lake Jackson and Galveston similar to the way it explains



City of El Paso Log of Consumption per Household





City of Abilene Log of Consumption per Household

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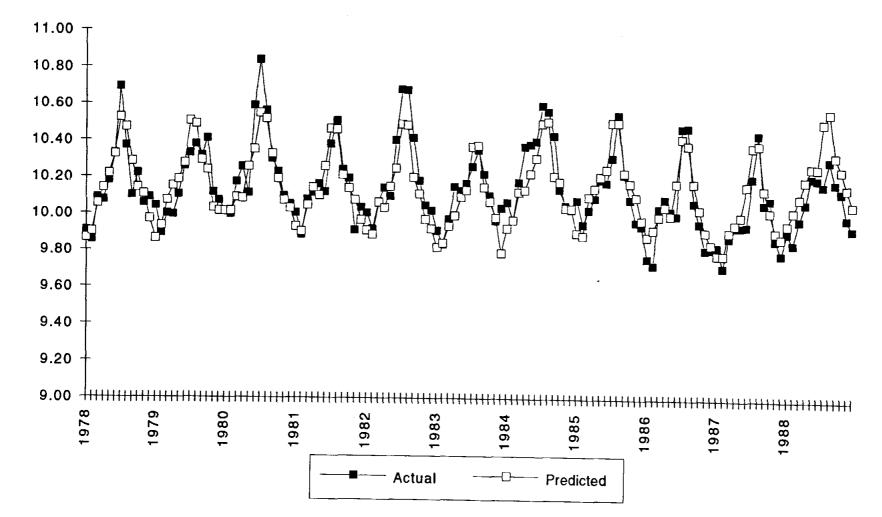
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City of San Antonio Log of Consumption per Household



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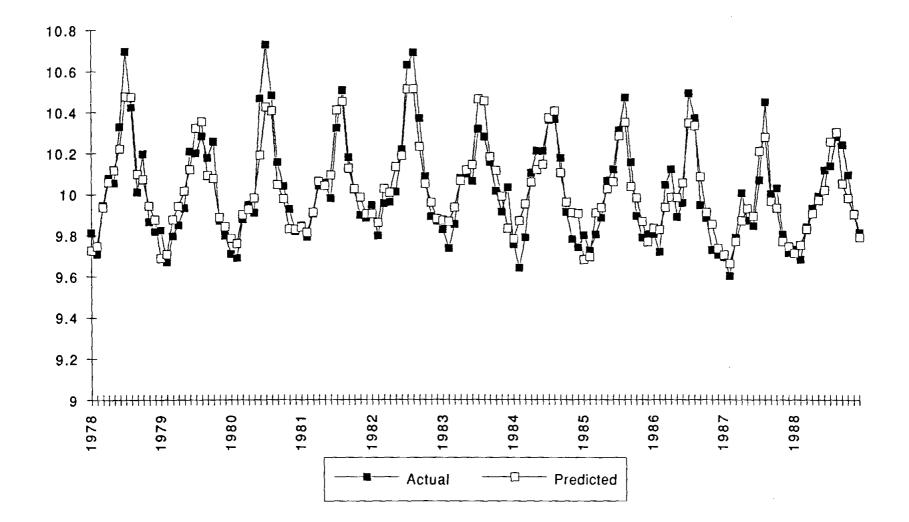
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City of Austin Log of Consumption per Household



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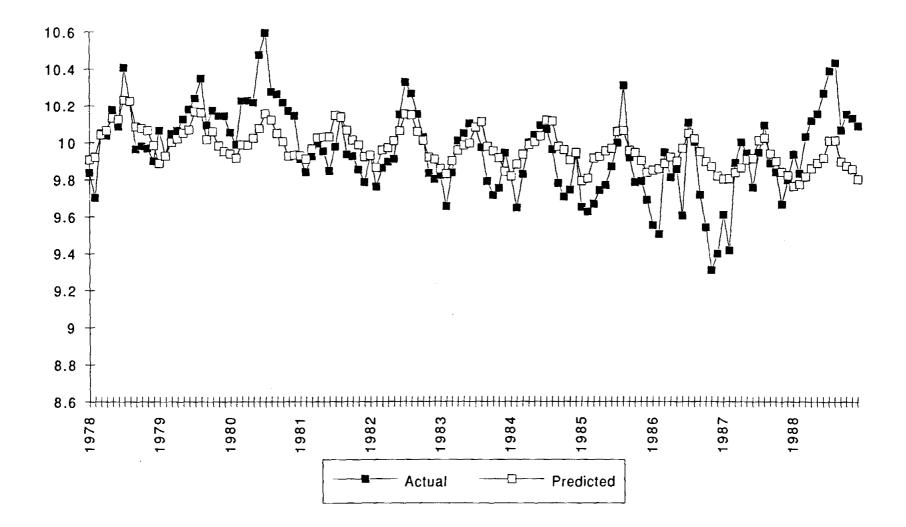
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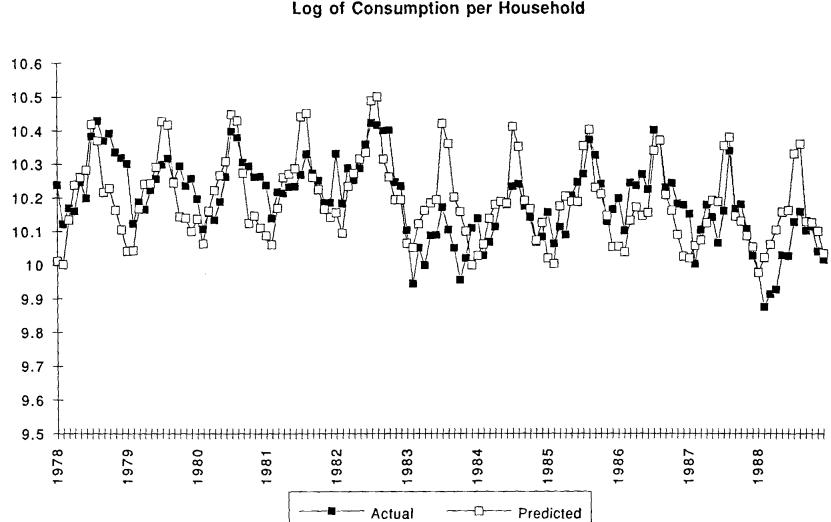
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City of Corpus Christi Log of Consumption per Household





City of Houston Log of Consumption per Household

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Figure V-7

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actual consumption for all three cities, although Lake Jackson and Galveston had no conservation programs until as recently as 1988.

There is no constant pattern of overestimation at El Paso and Austin, two other Texas cities with notable conservation programs implemented in the mid 1980s. If overestimation in the late 1980s could be taken as an indication of the model's inability to capture the effects of conservation, residuals should show a positive upward trend. That is, the residuals would reflect this failure and show rising, positive values, but they do not (**Appendix B**).

In summary, the statistical tests of the models give us confidence that we have explained a large portion of the variation in monthly consumption by the set of variables that we expect to be important, namely price, income, commercial concentration, dry days, temperature and city/regional location. Further, we have confidence that while some equations show non-uniform patterns of actual minus predicted consumption, such patterns do not seem consistent with expected conservation program effects. While conservation programs are no doubt having some impact in certain cities, we are unable to quantify such with econometric methods, given the available data.

#### Elasticity Estimates

Price elasticity of demand measures the response of consumption to a one-percent change in the price of water. **Table V-3** shows summer and winter price elasticities for each region. The highest elasticity coefficients are for MSAs in the East region. The coefficients are lowest in the Valley region, indicating that consumption will drop only slightly as price rises. Water is a more precious commodity in the Valley than in the Rolling Plains. Price elasticity increases slightly for non-peak consumption periods except in Victoria (I-35), Dallas/Fort Worth (Metroplex), Sherman/Denison and Waco, where elasticity decreases in nonpeak consumption periods, and in the East region which has constant elasticity coefficients due to the log-log form of the equation.

The response of consumption to a one-percent change in income is quite variable over the state (**Table V-4**). The response to rising income is lowest in the Valley region, at 0.031 in the McAllen-Edinburg-Mission MSA. The highest income elasticities are in MSAs in the Rolling Plains region, where Wichita Falls area residents tend to increase water consumption in winter by 2.3 percent for every onepercent increase in income. TABLE V-3. PRICE ELASTICITY OF DEMAND FOR WATER BY REGION BY MSA IN TEXAS

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	MSAIREGION	EAST	1-35	METRO- PLEX	METRO- PLEX SUB	ROLLING PLAINS	WEST	SOUTH- EAST	CENTRAL	VALLEY
	ABILENE					-0.173 -0.450				
	AMARILLO					0.400	-0.047 -0.048			
	AUSTIN		-0.293 -0.316				-0.048			
	BEAUPORT	-0.090	-0.316							
<b>B</b> .A.	BRAZORIA	-0.090						-0.087		
	BRNSVHAR							-0.088		-0.024
	BRYANCOL								-0.127	-0.026
_	CORPUS							-0.074	-0.130	
	DALLAS			-0.066	-0.177			-0.078		
-	EL PASO			-0.065	-0.187		-0.042			
	FTWORTH			-0.066	-0.177		-0.043			
	GALVESTON			-0.065	-0.187			-0.108		
	HOUSTON	-0.090						-0.110		
	KILLTEMP	-0.090							-0.143	
	LAREDO					-0.095			-0.147	
	LONGMARS	-0.090				-0.159				
	LUBBOCK	-0.090					-0.068			
	MCAEDMIS						-0.071			-0.033
-	MIDLAND						-0.075			-0.034
	ODESSA						-0.078 -0.072			
	SANGELO					-0.167	-0.078			
	SANTONIO		-0.224			-0.288				
	SHERMDEN		-0.228						-0.132	
	TEXARKNA	-0.090							-0.130	
	TYLER	-0.090 -0.090								
	VICTORIA	-0.090	-0.216							
	WACO		-0.207						-0.134	
	WICHITAF					-0.181			-0.133	
				set is the v	value for	-0.543	4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4	nd in face		

NOTE: The first elasticity in each set is the value for summer and the second is for winter.

TABLE V-4. INCOME ELASTICITY OF DEMAND FOR WATER BY REGION BY MSA IN TEXAS

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	MSA\REGION	EAST	1-35		METRO- PLEX SUB	POLLING PLAINS	WEST	SOUTH-	CENTRAL	VALLEY
	ABILENE	[		i tester X		0.818 2.056		<u> </u>		
	AMARILLO					2.000	0.205 0.205			
l-ring	AUSTIN		0.941				0.205			
	BEAUPORT	0.533	0.941							
	BRAZORIA	0.533						0.738 0.738		
	BRNSVHAR	8 1						0.730		0.034 0.035
	BRYANCOL								1.267	0.035
	CORPUS							0.712 0.712	1.267	
	DALLAS	8		0.802 0.802	0.962 0.962			0.712		
	EL PASO			0.002	0.962		0.194 0.194			
	FTWORTH			0.802	0.962		0,194			
-	GALVESTON			0.802	0.962			0.733		
	HOUSTON	0.533						0.733		
	KILLTEMP	0.533							0.978	
	LAREDO					0.492			0.978	
	LONGMARS	0.533 0.533				0.812				
	LUBBOCK	0.533					0.244 0.244			-
	MCAEDMIS						0.244			0.031
-	MIDLAND						0.302			0.032
_	ODESSA						0.302 0.259 0.259			
	SANGELO					0.779 1.540	0.209			
	SANTONIO		0.948 0.948			1.540				
	SHERMDEN		0.340						1.006 1.006	
******	TEXARKNA	0.533 0.533							1.000	
	TYLER	0.533 0.533								
-	VICTORIA	0.555	0.840 0.840							
	WACO		0.040						1.015 1.015	
	WICHITAF					0.836			1.010	
	NOTE: The fi	rst elasticit	y in each	set is the v	alue for su	2.301 mmer and	the secor	nd is for v	vinter.	}

#### VI. USE OF ECONOMETRIC MODELS FOR FORECASTING

This section of the report provides an example of how to use the econometric models for forecasting per capita monthly water consumption. The forecasts presented here are for exemplary purposes only. A forecast for planning use should be done paying particular attention to reasonable projections of region and city specific independent variables. The common set of assumed projections of independent variables used here are, however, within a reasonable range for the examples chosen.

#### Projections of Independent Variables

The independent variables which determine the following trends of per capita water consumption in Texas are price and income. All nine regional models include FAMLYINC, described as income per household which is the product of real per capita income and the number of persons per residential connection. Four scenarios for price and income were used to derive four alternative forecasts of water consumption for El Paso, Abilene, Dallas, San Antonio, Corpus Christi and Houston. The assumptions for marginal price (MP) in the forecast period were 1) flat real prices throughout the forecast period 1989-1999, using the monthly values for 1988, the last year of historic data; and 2) annual growth in real prices of 4.1% (the average rate of increase for the 72 cities during 1980-1988). The assumptions for per capita income were 1) annual growth of 1.5% over the forecast period (the Texas Comptroller's current 20 year forecast rate of increase), and 2) zero growth, keeping income unchanged from the December 1988 level.

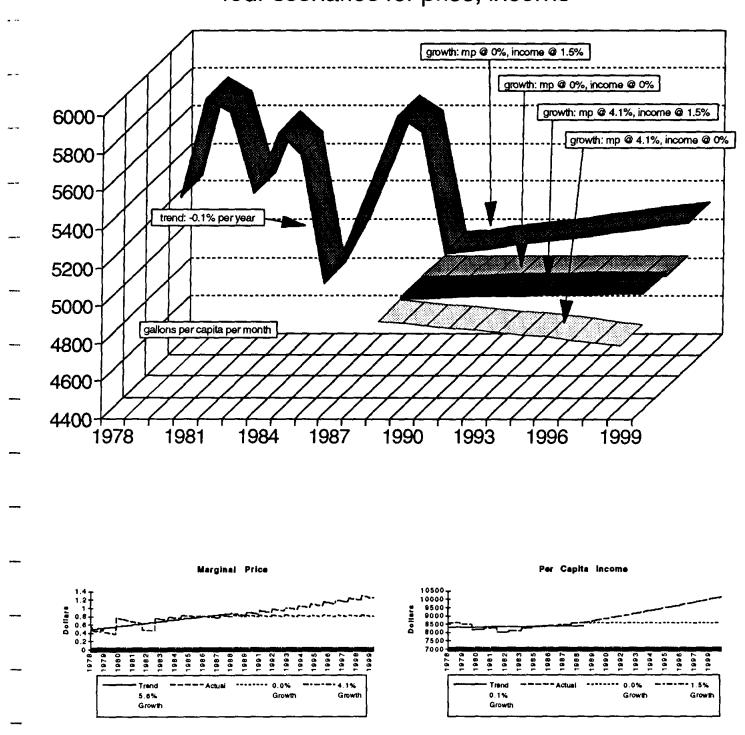
Persons per residential connection, and the commercial growth variable (COMPROXY) were held constant at their December 1988 levels. The number of dry days (DAYS) and the average monthly temperatures (TEMP) were forecast by projecting the average monthly values for the historic period.

#### Projections of Per Capita Water Consumption

Forecasts of per capita water consumption for El Paso, Abilene, Dallas, San Antonio, Austin, Corpus and Houston are shown with projected assumptions of price, income and persons per residential connection in **Figures VI-1 and VI-7**. Historical data are also shown for perspective. The forecast results are summarized below by price/income scenario:

a. <u>Growth rates of 0% for price, 1.5% for personal</u> <u>income</u>. Water consumption forecasts for all seven

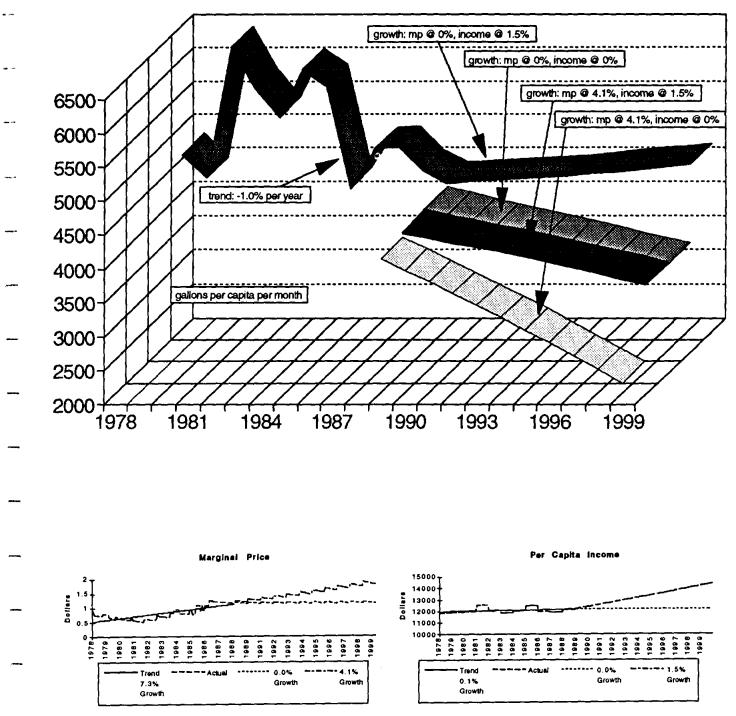
### EL PASO WATER CONSUMPTION four scenarios for price, income





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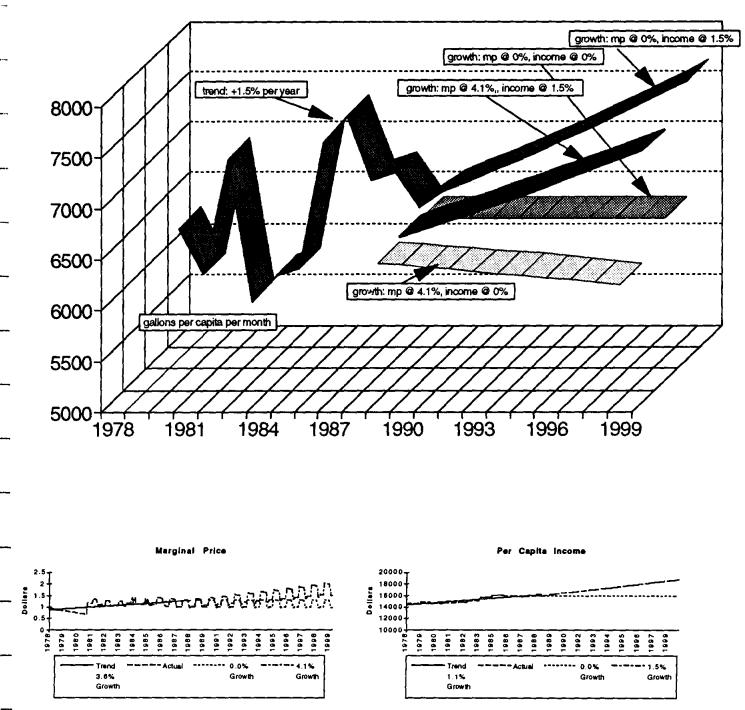
### ABILENE WATER CONSUMPTION four scenarios for price, income

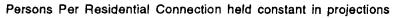


Persons Per Residential Connection held constant in projections

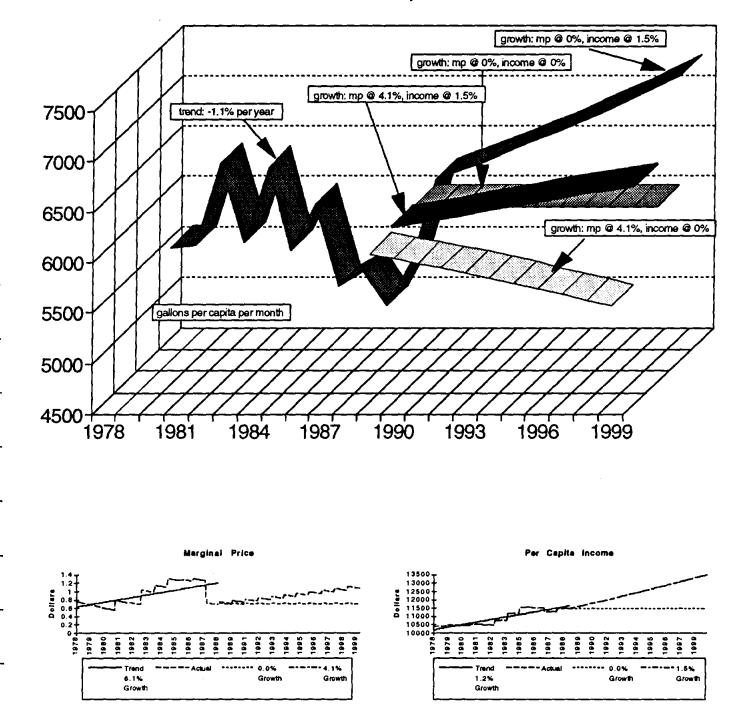
# DALLAS WATER CONSUMPTION

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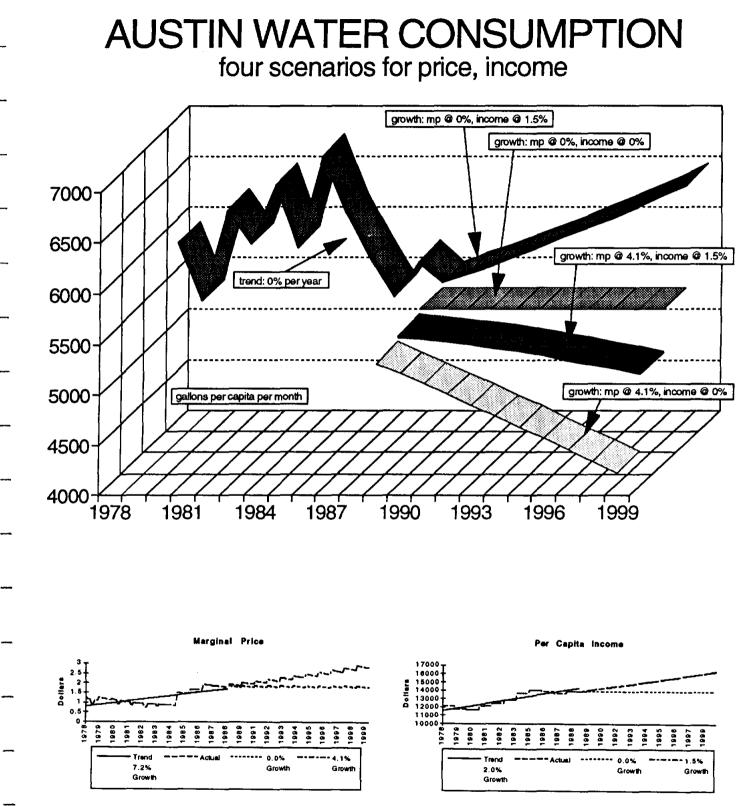


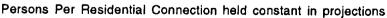


### SAN ANTONIO WATER CONSUMPTION four scenarios for price, income



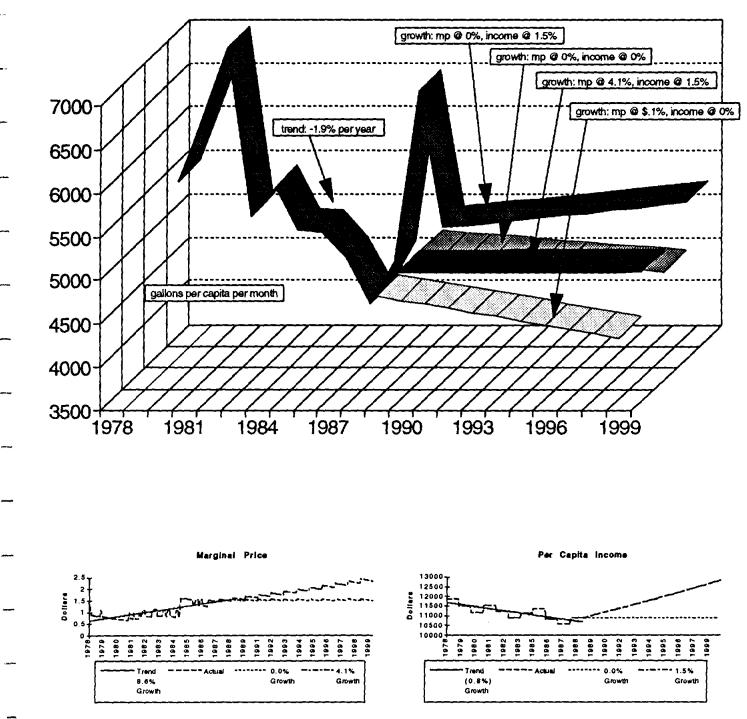
Persons Per Residential Connection held constant in projections





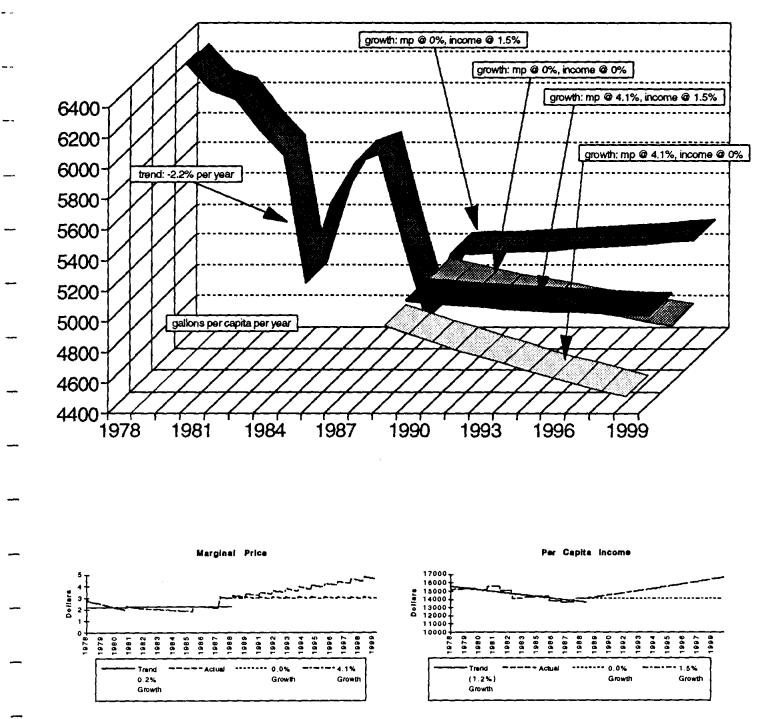
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### CORPUS WATER CONSUMPTION four scenarios for price, income



Persons Per Residential Connection held constant in projections

### HOUSTON WATER CONSUMPTION four scenarios for price, income



Persons Per Residential Connection held constant in projections

cities is projected to rise over the 1989-1999 period. Especially sharp increases are shown for Dallas, San Antonio and Austin relative to the other cities. The level of persons per residential connection for all these cities fell from 1978 to 1988. The arresting of that downward trend with a flat-growth assumption boosts the FAMLYINC variable for each city and therefore, the level of consumption. Persons per residential connection trended higher or remained virtually stable through 1988 for Houston, El Paso, Corpus and Abilene. The flat-growth assumption had a dampening effect, if any on income in the regional equations.

Note from **Table V-4** that the income elasticities of demand for water for the Dallas, San Antonio, Austin and Houston MSAs are higher than for the El Paso, Abilene and Corpus MSAs, and higher than most MSAs in the table. Income elasticity of demand measures the response of consumption to a one-percent change in income per residential connection. With other factors held constant, a 1.5% increase in income at San Antonio will yield an increase in water consumption of 1.4%. A 1.5% increase in income for El Paso yields only a 03% increase in water consumption, and **Figure VI-1** reflects the smaller response to income change.

- b. <u>Growth rates of 0% for price and 0% for personal</u> <u>income</u>. This scenario is unlikely over the period 1989-1999. This scenario produces forecasts for El Paso, Dallas, San Antonio and Austin of zero-growth in consumption because income and price are held constant, along with other predictive variables in the forecast period. Only for the cities of Abilene, Corpus and Houston, whose equations include the variable TIME, does consumption change over the forecast period. Consumption declines for Abilene, Corpus and Houston because TIME is inversely related to water consumption in their respective regional models.
- c. <u>Growth rates of 4.1% for price and 1.5% for personal</u> <u>income</u>. This scenario is very plausible. Under these assumptions of price and income growth, per capita water consumption would increase through the forecast period for El Paso, Dallas and San Antonio; would decrease for Abilene, Austin and Houston; and would remain virtually unchanged for Corpus.

The effect of the assumption of flat-growth in persons per residential connection is important again as it was in scenario a, but its importance is countered by the relative importance of price growth as seen by comparing price elasticities in **Table V-3** with income elasticities in Table V-4.

Price elasticity of demand for water is a measure of the percent change in water consumption associated with a one-percent change in the price of water. Dallas consumption under scenario "a" shows an upward growth rate of 1.6% annually in Figure VI-3, assuming 1.5% growth in income and 0% growth in price. With the assumption of 4.1% annual growth in price, consumption still shows upward growth over the forecast period. This persistent growth in consumption for Dallas is reflected in a relative insensitivity to water price changes combined with a relatively high income elasticity. The price elasticity for Dallas is quantified as -0.066. Of the MSAs representing the forecast cities, only the El Paso MSA has a lower price elasticity than that of Dallas (Table V-3). Price elasticity for Austin is nearly 4.5 times greater than for Dallas. The result of a 4.1% increase in price for Austin would result in declining consumption over the forecast period, other factors held constant.

d. <u>Growth rates of 4.1% for price and 0% for personal</u> <u>income</u>. It is perhaps unlikely to have escalating real prices occurring with flat real personal income in the forecast period, but it is possible since that is basically the experience of the 1980s. If such a scenario were to happen, consumption would decline in all seven forecast cities.

#### Conservation Program Adjustments to Forecasts

Attempts to quantify conservation program effects using econometric methods applied in this analysis were not successful. Either the data are too weak, the effects not yet evident or the effects are not very important, independent of the other variables included in the models. One difficulty is the non-uniform definition of what constitutes a conservation program. The most important ambiguity is probably pricing. The analysis here reported includes marginal prices and, in many cases, utilities switched from flat or declining block rate structures to increasing block structures during the period of analysis, and classify such a change as a conservation program. If one accepts such a definitive, then the analysis in this report quantifies such a relationship.

The effects of mandatory government rules concerning appliance standards are not explicitly included in the current analysis. Projections of the effects of such mandates may be included for planning purposes by subtracting expected impacts from forecasts made using the

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equations estimated in this study, but such a practice is an ad hoc method that is apt to overstate the case. That is, one would not expect per capita consumption reductions forthcoming from higher prices to be the same response that would ensue after water saving devices have been installed under government rules. The problem is that we cannot be sure how much reliance to place on the elasticity estimates in a market where behavior has been changed by government rule when the estimates of price response came for a market where such rules did not exist. The other difficulty is that municipal water is supplied by cost of service regulated utilities who may change rates in the future in a different time path under mandtory appliance standard than would be the case without them. The point is that price and price elasticities may be different with and without mandatory appliance standards so that accounting for the impacts of one cannot be considered independent of the other. This topic may need further research.

#### VII. SUMMARY AND FINDINGS

#### Summary

Eleven years of monthly consumption data for each of 72 cities in 28 MSAs of Texas were analyzed in this study in order to determine the underlying causes of declining per capita water use. Nine regional econometric models were estimated by grouping the 28 MSA sets of cities into homogeneous climatic and geographical groups. Each model allows an explanation of historical water consumption for each city in the group.

The set of six regional models all contain variables that we expect a priori to be important determinants of per capita water consumption. The equation forms and specific variables included differ among regions, and in some cases variables were ultimately dropped from the final equation because the estimated parameters were statistically insignificant; that is, we could not say with confidence that they had anything at all to do with consumption. In the end, however, the variables we believe should explain water consumption do in fact test significant and include (1) marginal price, (2) household income, (3) number of dry
(low rainfall) days in the month, (4) temperature and (5) the concentration of commercial customers on the system. The fact that different forms and model specifications apply to different regions of the state also means that regional location and city size are important in explaining per capita municipal water consumption. The statistical properties of the models are all quite acceptable, and in fact are improved over many such results cited in the literature.

Price elasticity of demand estimates from the 72 cities in nine regions range from -0.042 to -0.543 while income elasticity estimates range from 0.031 to 1.267. These elasticities are well within the range of estimates obtained by others in the econometrics field. These elasticities allow simple calculation of the expected demand response to price and income changes.

#### Findings

The study of per capita municipal water demand during the 1978-1988 period leads to some interesting and important findings. The first important finding is that price, household income, concentration of commercial users, weather conditions, city size and location are all important variables in explaining historical water consumption and for forecasting future consumption. Seasonal variations are mostly explained by temperature and the lack of rainfall.

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Long term trends are explained by household income, price and concentration of commercial users.

The general downward trend in per capita water consumption during 1978-1988 was the result of two sets of forces working at different parts of the time period, but themselves interrelated. The late 1970s and early 1980s brought rapid economic growth to Texas cities, resulting in at one and the same time, higher per capita incomes (exceeding the national average in 1982 for the first time ever) and explosive growth. Municipalities responded by constructing new facilities planned to catch up with growth and to meet a continued high growth in demand. By the mid-1980s growth had stopped and debt service requirements began to be realized, forcing utility rates to rise. Water supply, treatment and wastewater disposal costs also increased due to growing scarcity of supply and more stringent wastewater regulations. During this period of rising rates, many utilities switched from flat or declining block rate structures to increasing block structures, meaning that the marginal price of water rose above the average cost, a reversal of the historical relationship of the two prices. This sequence of events - rapid income and population growth - followed by stagnation and the lagged supply price response by cost of service based utilities meant that consumers were hit with stagnating incomes and rising marginal prices of water at the same time. The net result was a decline in per capita consumption rates.

If the above explanation of the past eleven years is correct, the question arises, "Will this downward trend in per capita consumption continue?" The analysis suggests that if the same forces of price, income, weather and persons per connection continue to determine consumption, per capita consumption is likely to continue declining in the foreseeable future.

There are two reasons why the trend, as influenced by the above factors, will continue. First, although Texas is now coming out of the longest recession it has had since WWII, the long term prospects for a per capita income growth near that of the post WWII era will be difficult to attain. Second, the overbuilding of utility capacity which occurred in the 1980s, leaving us with considerable excess capacity and high prices, should begin to abate soon, perhaps relieving the upward price pressure for a time, but in the long term prices are destined to rise rapidly. Most of the real price increases needed to retire the debt of the overbuilding have already been realized and real rates should begin to decline. The net result is that per capita consumption is likely to decline or stabilize in the long term.

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The above conclusion will be reinforced by public policy driven by a number of interests ranging from public finance to environmental concerns. A case in point is an initiative to require certain water saving technologies to be installed by users, perhaps with the help of a public subsidy. Such mandates could further alter the consumption levels and trends of the future.

#### APPENDIX A

Statistical Output of Six Regional Models

#### Model: CENTRAL REGION: LOG-LINEAR Dependent Variable: LOGCONS

			Anal	ysis of	Varian	ce		
	Source			Sum of Juares		Mean uare	F Valu	e Prob>F
# -=	Model Error C Total		1305 91.	15052 88586 03638		6789 7041	182.75	5 0.0001
		t MSE Mean	0.26535 9.68666 2.73933	i Adj	quare R-sq	0.6		
			Para	meter Es	timate	s		
	Variable	DF	Parameter Estimate		dard rror	T for H( Parameter	-	Prob >  T
-	INTERCEP MP	1	8.070488 -0.121359	0.1309 0.0381	5518	61.6 -3.1	181	0.0000 0.0015
	FAMLYINC TEMP DAYS COMPROXY	1 1 1	0.000028325 0.009232 0.010401 0.758233	0.0000 0.0006 0.0032 0.2481	3210 6716	14.9 14.6 3.1 3.0	506 183	0.0001 0.0001 0.0015 0.0023
	BRY BEL WAC	1 1 1	-0.252587 -0.348467 0.363213	0.0390 0.0439 0.0330	5116 2496	-6.4 -7.9 10.9	168 933	0.0001 0.0001 0.0001
	WOO COP KIL	1 1 1	0.176476 -0.385938 -0.531600	0.0337 0.0433 0.0292	6189 6887	5.2 -8.8 -18.1	227 199	0.0001 0.0001 0.0001
	SHE DEN	1 1	-0.383254 -0.352713	0.0336 0.0351	9371 1444	-11.3 -10.0	175 )45	0.0001 0.0001
	DSEAS	1	0.179495	0.0251	7266	7.1	.31	0.0001

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#### Model: CENTRAL REGION: LOG-LOG Dependent Variable: LOGCONS

#### Analysis of Variance

Source		um of lares S	Mean quare F	Value Prob>F
Model Error C Total	14 177.9 1305 94.1 1319 272.0	L1250 O.	70885 170 07212	6.226 0.0001
Root MSE Dep Mean C.V.	0.26855 9.68666 2.77233	R-square Adj R-sq		
	Paran	neter Estimat	es	
Variable DF	Parameter Estimate	Standard Error	T for H0: Parameter=0	Prob >  T
INTERCEP1LOGMP1LOGFINC1LOGTEMP1LOGDAYS1BRY1BEL1WAC1WOO1COP1KIL1SHE1DEN1DSEAS1	$\begin{array}{c} -3.419455 \\ -0.160874 \\ 1.037026 \\ 0.510024 \\ 0.187562 \\ 0.090974 \\ -0.322024 \\ -0.332745 \\ 0.352985 \\ 0.217403 \\ -0.424992 \\ -0.539542 \\ -0.435916 \\ -0.401759 \\ 0.222774 \end{array}$	0.95563867 0.03960477 0.08182628 0.03773387 0.07605561 0.02269853 0.04064294 0.03770657 0.03332756 0.03512248 0.04810884 0.02973162 0.03466293 0.03656183 0.02422265	-3.578 -4.062 12.674 13.516 2.466 4.008 -7.923 -8.825 10.591 6.190 -8.834 -18.147 -12.576 -10.988 9.197	0.0004 0.0001 0.0001 0.0138 0.0001 0.0001 0.0001 0.0001 0.0001 0.0001 0.0001 0.0001 0.0001 0.0001 0.0001 0.0001 0.0001

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#### Model: CENTRAL REGION: LINEAR Dependent Variable: CONS

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#### Analysis of Variance

Source	DF	Sum of Squares		F Value	Prob>F
Model Error C Total	1305 50		6131938568.2 38876322.131	157.729	0.0001
Root MSE Dep Mean C.V.	6235.0 18057.7 34.5	2424	R-square Adj R-sq	0.6285 0.6246	

#### Parameter Estimates

<b>**</b> **	Variable	DF	Parameter Estimate	Standard Error	T for HO: Parameter=0	Prob >  T
	INTERCEP	1	-21392	3076.9268493	-6.952	0.0001
	MP	1	-2094.880831	896.55505433	-2.337	0.0196
	FAMLYINC	1	0.841070	0.04576768	18.377	0.0001
	TEMP	1	158.520521	14.85293142	10.673	0.0001
	DAYS	1	188.335911	76.77035019	2.453	0.0143
	COMPROXY	1	325.211345	5830.0878418	0.056	0.9555
	BRY	1	-6260.006893	917.60844340	-6.822	0.0001
	BEL	l	-4441.321867	1032.1311749	-4.303	0.0001
	WAC	l	5503.321598	777.41604396	7.079	0.0001
	WOO	1	4455.657161	793.32327048	5.616	0.0001
	COP	1	-5169.700281	1019.0643732	-5.073	0.0001
	KIL	1	-12312	687.99309407	-17.896	0.0001
	SHE	1	-7779.144144	791.72125551	-9.826	0.0001
	DEN	1	-6812.752594	825.10503953	-8.257	0.0001
	DSEAS	l	4504.858248	591.49713271	7.616	0.0001

#### CORRELATION ANALYSIS

	5 'V	AR' Variables:	MP	FAMLYINC	TEMP I	DAYS CO	MPROXY
			Simple	• Statistics	5		
	Variable		N	Mean	Std De	ev	Sum
	MP FAMLYINC TEMP DAYS	132 132 132 132	0 0 6 0 2	1.11217 36966 55.69037 27.12879	0.2365 620 14.4812 2.3432	01 48 27 20	1468 3794707 86711 35810
and the second s	COMPROXY	132 Simple Stati		0.09497	0.0599	94 125	5.35589
<b></b>	Variable	Minimu	m	Maximum			
	MP FAMLYINC TEMP DAYS COMPROXY	0.3384 2504 30.8700 5.0000 0.0100	2 0 8 0 3	2.07254 54936 9.23000 1.00000 0.30000			
Pea	arson Corn	elation Coeff	icients /	Prob >  R	under Ho:	Rho=0 / N	1 = 1320
1994		MP	FAMLYINC	I I	EMP	DAYS	COMPROXY
MP		1.00000 0.0	0.23771 0.0001		572 683	0.00671 0.8076	-0.20257 0.0001
FAMLY	YINC	0.23771 0.0001	1.00000 0.0		296 544	0.06463 0.0189	0.22793 0.0001
⊥EMP		-0.01572 0.5683	0.05296 0.0544			0.07000 0.0110	0.02417 0.3803
4ys		0.00671 0.8076	0.06463 0.0189	0.07 0.0	000 110	1.00000 0.0	-0.02355 0.3926
 DMPF	ROXY	-0.20257 0.0001	0.22793 0.0001		417 - 803	0.02355 0.3926	1.00000 0.0

y			Ana	lysis of V	ariance		
	Source		DF S	Sum of quares	Mean Square	F Val	lue Prob>F
# <b>*</b>	Model Error C Total		1701 19735	708524 377 785576 116 494100		325.7	748 0.0000
<b></b>		Mear Mear		l Adj		0.7283 0.7261	
			Par	ameter Est	imates		
antin.	Variable	DF	Parameter Estimate	Stand Er		for HO: meter=0	Prob >  T
_	INTERCEP TIME	1 1	-1579.937997 -0.364233	1124.6315 0.07521	000	-1.405 -4.843	0.1602 0.0001
<b>8</b> 000m	MP FAMLYINC TEMP DAYS	1 1 1 1	-1150.623861 0.129296 109.997310 112.519242	253.97766 0.01038 7.55527 28.58107	690 808	-4.530 12.448 14.559 3.937	0.0001 0.0001 0.0001 0.0001
-	COMPROXY D25 D4	1 1 1	20658 5877.016682 2154.868306	1650.2929 354.01056 274.49483	431 669	12.518 16.601 7.850	0.0001 0.0001 0.0001
-	TXK CON HOU BAY	1 1 1 1	8736.179747 9358.399530 12702 1408.586397	428.44861 395.11488 508.29071 370.94263	753 863 316	20.390 23.685 24.991 3.797	0.0001 0.0001 0.0001 0.0002
-	D16 DSEAS	1 1	4286.580079 2597.416524	277.69934 269.67593	581	15.436 9.632	0.0001 0.0001

#### Model: EAST REGION: LOG-LINEAR Dependent Variable: LOGCONS

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#### Analysis of Variance

Source	DF	Sum c Square		F Value	Prob>F
Model Error C Total	14 1701 1715	189.0242 67.8084 256.8326	6 0.03986	338.696	0.0000
Root MSE Dep Mean C.V.	9	.19966 .62164 .07511	R-square Adj R-sq	0.7360 0.7338	

#### Parameter Estimates

	Variable	DF	Parameter Estimate	Standard Error	T for HO: Parameter=0	Prob >  T
	INTERCEP	1	8.316178	0.06592118	126.153	0.0000
	TIME	1	-0.000020700	0.00000441	-4.696	0.0001
	MP	1	-0.054906	0.01488710	-3.688	0.0002
	FAMLYINC	1	0.000011725	0.00000061	19.258	0.0001
	TEMP	1	0.007283	0.00044286	16.445	0.0001
	DAYS	1	0.005839	0.00167530	3.485	0.0005
	COMPROXY	1	0.996068	0.09673325	10.297	0.0001
and the second	D25	1	0.394668	0.02075061	19.020	0.0001
	D4	1	0.225046	0.01608974	13.987	0.0001
	TXK	l	0.671280	0.02511386	26.729	0.0001
*****	CON	1	0.556300	0.02315998	24.020	0.0001
	HOU	1	0.604033	0.02979387	20.274	0.0001
	BAY	1	0.132475	0.02174310	6.093	0.0001
	D16	1	0.365347	0.01627757	22.445	0.0001
	DSEAS	1	0.127176	0.01580727	8.045	0.0001

#### Model: EAST REGION: LOG-LOG Dependent Variable: LOGCONS

• **e				Analysis d	of Varian	ice		
mj - 13	Source		DF	Sum of Squares		Mean Juare	F Value	Prob>F
<b>100</b> - 1	Model Error C Total		1692	189.06908 67.59957 256.66865		0493 3995	338.025	0.0000
1000 - 10		ot MSE Mean	9.6		l-square dj R-sq	0.73 0.73		
				Parameter	_			
	Variable	DF	Paramet Estima		andard Error	T for HO Parameter		) >  T
siliperanas,	INTERCEP LOGTIME	1 1	3.1322 -0.2113	28 0.03	267635 1766911	6.6 -5.6	10	0.0001 0.0001
	LOGMP LOGFINC LOGTEMP	1 1 1	-0.0901 0.5332 0.4192	79 0.02	2059166 2595968 2636114	-4.3 20.5 15.9	43	0.0001 0.0001 0.0001
-	LOGPROXY LOGDAYS D25	1 1 1	0.0691 0.2662 0.3775	85 0.05 79 0.02	941798 493009 088575	7.3 4.8 18.0	48 78	0.0001 0.0001 0.0001
_	D4 TXK CON HOU	1 1 1 1	0.1818 0.7718 0.5644 0.6132	95 0.02 78 0.02	634712 829364 305658 761729	11.1 27.2 24.4 22.2	82 82	0.0001 0.0001 0.0001 0.0001
~	BAY D16 DSEAS	1 1 1	0.1078	74 0.02 30 0.01	177453 575022 527212	22.2 4.9 23.5 9.5	54 57	0.0001 0.0001 0.0001

EAST REGION

#### CORRELATION ANALYSIS

	6 'VAR' Variables:	LOGTIME	LOGMP 1	LOGFINC	LOGTEMP	LOGDAYS	LOGPROXY
1 9.46.4		Si	mple Statis	stics			
(1-1-1- <b>1</b> -1-1-1-1-1-1-1-1-1-1-1-1-1-1-1-	Variable	N	Mea	ı	Std Dev		Sum
	LOGTIME LOGMP	1716 1716	0.2591 0.2591	-	0.31671 0.31671		63488 63488
() - 44 <b>m</b>	LOGFINC LOGTEMP	1716 1716	10.60310		0.35252 0.22288		18195 7142
	LOGDAYS LOGPROXY	1707 1716	3.2659	7	0.08990		5575 -4215
	-			-			

#### Simple Statistics

Nam y Connega	Variable	Minimum	Maximum
	LOGTIME LOGMP LOGFINC LOGTEMP LOGDAYS LOGPROXY	-0.54000 -0.54000 9.58014 3.45774 2.89037 -3.90197	1.13051 1.13051 11.30459 4.46740 3.43399 -0.98450

Pearson Correlation Coefficients / Prob > |R| under Ho: Rho=0 / Number of Observations

Mining	LOGTIME	LOGMP	LOGFINC	LOGTEMP	LOGDAYS	LOGPROXY
LOGTIME	1.00000	1.00000	0.08991	-0.02804	0.02123	-0.26653
	0.0	0.0	0.0002	0.2456	0.3808	0.0001
	1716	1716	1716	1716	1707	1716
LOGMP	1.00000	1.00000	0.08991	-0.02804	0.02123	-0.26653
	0.0	0.0	0.0002	0.2456	0.3808	0.0001
	1716	1716	1716	1716	1707	1716
LOGFINC	0.08991	0.08991	1.00000	0.16801	0.04395	0.22642
	0.0002	0.0002	0.0	0.0001	0.0695	0.0001
	1716	1716	1716	1716	1707	1716
- Logtemp	-0.02804	-0.02804	0.16801	1.00000	0.12853	0.05454
	0.2456	0.2456	0.0001	0.0	0.0001	0.0239
	1716	1716	1716	1716	1707	1716
LOGDAYS	0.02123	0.02123	0.04395	0.12853	1.00000	0.00125
	0.3808	0.3808	0.0695	0.0001	0.0	0.9589
	1707	1707	1707	1707	1707	1707
LOGPROXY	-0.26653	-0.26653	0.22642	0.05454	0.00125	1.00000
	0.0001	0.0001	0.0001	0.0239	0.9589	0.0
	1716	1716	1716	1716	1707	1716

Model: I-35 REGION: LINEAR Dependent Variable: CONS

MP

TEMP

DAYS

D3

D22

DSEAS

FAMLYINC

ç <b>**</b>	Analysis of Variance							
	Source		Sum of quares S	Mean quare FV	/alue Prob>F			
	Model Error C Total	388 343306	634292 2176090 8803.4 8848115 703095		5.938 0.0001			
<b></b> .	Root MSE Dep Mean C.V.	2974.5782 21774.6717 13.6607	2 Adj <sup>R</sup> -sq					
		Par	ameter Estimat	es				
	Variable DF	Parameter Estimate	Standard Error	T for HO: Parameter=0	Prob >  T			
	INTERCEP 1	-20263	3215.9384331	-6.301	0.0001			

620.15135631

0.05434269

14.56079069

71.40199608

471.03971698

493.00895467

469.20350222

-8.972

6.347

15.626

4.618

15.756

17.625

11.053

0.0001

0.0001

0.0001

0.0001

0.0001

0.0001

0.0001

-5563.864316

0.344895

227.528724

329.728042

7392.749386

8301.879576

5449.013913

1

1

1

1

1

1

1

#### Model: I-35 REGION: LOG-LINEAR Dependent Variable: LOGCONS

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### Analysis of Variance

Source	DF	Sum o Square		lean lare	F Value	Prob>F
Model	7	30.7304	5 4.39	006	332.095	0.0001
Error	388	5.1290	9 0.01	322		
C Total	395	35.8595	4			
Root MSE	0	.11498	R-square	(	0.8570	
Dep Mean	9	.94242	Adj <sup>®</sup> R-sq	(	0.8544	
c.v.	1	.15641				

#### Parameter Estimates

 Variable	DF	Parameter Estimate	Standard Error	T for H0: Parameter=0	Prob >  T
INTERCEP	1	7.964865	0.12430443	64.075	0.0001
MP	1	-0.244931	0.02397047	-10.218	0.0001
 FAMLYINC	1	0.000018551	0.00000210	8.832	0.0001
TEMP	1	0.010514	0.00056281	18.682	0.0001
DAYS	1	0.011598	0.00275987	4.202	0.0001
D3	1	0.352806	0.01813594	19.453	0.0001
 D22	1	0.388082	0.01820692	21.315	0.0001
DSEAS	1	0.187767	0.01905609	9.853	0.0001

Model: I-35 REGION: LOG-LOG Dependent Variable: LOGCONS

	Analysis of Variance								
an sea	Source			Sum of quares		Mean Jare	F Val	lue	Prob>F
900 · **	Model Error C Total		388 5	.38941 .47013 .85954		4134 1410	307.9	935	0.0001
		t MSE Mean	0.1187 9.9424 1.1942	2 Adj	quare R-sq		8475 8447		
-			Par	ameter Es	timates	5			
-	Variable	DF	Parameter Estimate		ldard Irror	T for Paramet		Prob >	א  ד
	INTERCEP LOGMP LOGFINC LOGTEMP LOGDAYS D3 D22		-3.620332 -0.261124 0.894294 0.640729 0.279285 0.345736 0.382372	1.1629 0.0268 0.1047 0.0368 0.0743 0.0189 0.0191	9529 7027 7404 8467 4379	-9 8 17 3 18	1.113 .709 .536 .376 .755 .251 .019	0 0 0 0	.0020 .0001 .0001 .0001 .0002 .0001 .0001
	DSEAS	1 1	0.215919	0.0190			.332		0001

I-35 REGION

#### CORRELATION ANALYSIS

4 'VAR' Variables: MP FAMLYINC TEMP DAYS

#### Simple Statistics

Variable	N	Mean	Std Dev	Sum
MP	396	1.01047	0.32806	400.14477
FAMLYINC	432	48962	4015	21151478
TEMP	432	68.92685	12.65283	29776
DAYS	432	27.35880	2.12454	11819

#### Simple Statistics

990 - De	Variable	Minimum	Maximum
	MP	0.55944	1.94192
	FAMLYINC	41498	57675
	TEMP	40.39000	88.11000
	DAYS	19.00000	31.00000

### Pearson Correlation Coefficients / Prob > |R| under Ho: Rho=0 / Number of Observations

	MP	FAMLYINC	TEMP	DAYS
MP 	1.00000 0.0 396	0.51068 0.0001 396	-0.05152 0.3065 396	0.02505 0.6192 396
_FAMLYINC	0.51068	1.00000	-0.04084	0.05952
	0.0001	0.0	0.3972	0.2170
	396	432	432	432
¶EM₽	-0.05152	-0.04084	1.00000	-0.02557
	0.3065	0.3972	0.0	0.5961
	396	432	432	432
)AYS	0.02505	0.05952	-0.02557	1.00000
	0.6192	0.2170	0.5961	0.0
	396	432	432	432

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## Model: METROPLEX REGION: LINEAR Dependent Variable: CONS

.

•-	Source		DF S	Sum of Squares		Mean uare	F Val	ue	Prob>F
	Model Error				75461209 17395622		433.7	94	0.0001
	C Total	-	923 68753	7237221					
		ot MSE Mean		26 ž	R-square Adj R-sq		.7683 .7665		
_			Parameter		Estimate tandard	T for			
	Variable	DF	Estimate		Error	Parame	ter=0	Prob >	<b> T  </b>
	INTERCEP MP	1 1	-33339 -959.272337	413.05	3729505 5544855	-	2.913 2.322	0.0 0.0	204
	FAMLYINC TEMP DAYS	1 1 1	0.336998 281.819178 506.864726		3246403 )629529 )032717	2	0.381 4.493 7.712	0.0 0.0 0.0	001
	COMPROXY D11 DSEAS	1 1 1	39487 2539.022545 7058.573488	5795.5 370.14	5374962 1284720 1309413		6.813 6.860 4.230	0.0 0.0 0.0	001 001

#### Model: METROPLEX REGION: LOG-LINEAR Dependent Variable: LOGCONS

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		Anal	ysis of Varia	ance		
Source			um of	Mean	Value	Prob>F
Source		DF Sq	uares S	Square F	Value	PIOD>F
Model Error C Total		916 27.		.77645 4 .03038	86.338	0.0001
	t MSE Mean •		R-square Adj R-se			
		Para	meter Estimat	tes		
Variable	DF	Parameter Estimate	Standard Error	T for H0: Parameter=	0 Prob >	T
INTERCEP MP FAMLYINC TEMP DAYS COMPROXY D11 DSEAS		$\begin{array}{r} 7.315527 \\ -0.053469 \\ 0.000014511 \\ 0.014284 \\ 0.024448 \\ 2.137456 \\ 0.142162 \\ 0.201805 \end{array}$	0.10790235 0.01726253 0.00000136 0.00048087 0.00274660 0.24220871 0.01546911 0.02073077	67.79 -3.09 10.69 29.70 8.90 8.82 9.19 9.73	7       0.0         5       0.0         3       0.0         1       0.0         5       0.0         0       0.0	001 001 001 001
	-			2173		~ ~ ~

#### Model: METROPLEX REGION: LOG-LOG Dependent Variable: LOGCONS

a			Ana	alysis d	of Variar	nce			
				Sum of		Mean		_	_
• -7	Source		DF S	quares	Sq	quare	F Va.	lue	Prob>F
dar di	Model Error C Total		916 31	.60714 .65895 .26608		22959 )3456	411.	710	0.0001
<b>11.1</b> (A)		t MSE Mean	0.1859 9.9066 1.8766	0 A	l-square dj R-sq		0.7588 0.7570		
			Par	ameter	Estimate	s			
	Variable	DF	Parameter Estimate	st	andard Error	T for Parame	r HO: eter=0	Prob >	T
	INTERCEP LOGMP LOGFINC	1 1 1	-5.279767 -0.073329 0.929430	0.02	394871 324518 492974	-	-5.906	0.0	001
	LOGFINC LOGTEMP LOGDAYS LOGPROXY	1 1 1	0.792572 0.595566 0.105710	0.C2 0.07	492974 995773 601118 459278		L2.404 26.456 7.835 7.244	0.0	)001 )001 )001 )001
	D11 DSEAS	1 1	0.128456 0.270423	0.01	753912 097728	]	7.324 L2.891	0.0	0001

#### CORRELATION ANALYSIS

	5 'VA	R' Variables:	МР	FAMLYINC	TEMP	DAYS	COMPROXY			
••••••••••••••••••••••••••••••••••••	Simple Statistics									
<b>ana</b> i na ta	Variable	1	1	Mean	Std	Dev	Sum			
	MP FAMLYINC TEMP DAYS COMPROXY	924 924 924 924 924 924	1 65 1 27	.20908 55038 .27315 .46861 .08372	0.33 7 15.11 2.26 0.04	975 004 617	1117 50855321 60312 25381 77.35477			
Party and an	COMINOAT	Simple Statis			0.04	000	,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,			
-	Variable	- Minimun	n Ma	aximum						
	MP FAMLYINC TEMP DAYS COMPROXY	0.64000 39711 32.58500 15.00000 0.01000	90 90 91	.34005 76343 .90000 .00000 .20000						
Pe	arson Corre	elation Coeffi	.cients / 1	Prob >  R	under H	o: Rho=0 /	/ N = 924			
-		MP	FAMLYINC	Т	EMP	DAYS	COMPROXY			
MP —		1.00000 0.0	0.01219 0.7114	0.02 0.4	425 616	-0.04852 0.1405	0.10195 0.0019			
FAML	YINC	0.01219 0.7114	1.00000 0.0	-0.00 0.9	263 365	-0.00092 0.9777	0.76307 0.0001			
_EMP		0.02425 0.4616	-0.00263 0.9365	1.00 0.0		0.09801 0.0029	0.01196 0.7165			
AYS	-	-0.04852	-0.00092	0.09	801	1.00000	-0.03465			

0.9777

0.76307

0.0001

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OMPROXY

0.1405

0.10195

0.0019

A-17

0.0029

0.7165

0.01196

0.0

-0.03465

0.2927

0.2927

1.00000

0.0

#### Model: METROPLEX SURBURBAN: LINEAR >ependent Variable: CONS

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#### Analysis of Variance

Source		DF S	Sum of Squares	Mean Square	F Va]	ue Prob>F			
Model			787930 58735		411.3	0.0000			
Error				453.875					
C Total		1451 67595	482872						
Roo	t MSE	3778.9487	8 R-squa	ire	0.6951				
Dep	Mean	15463.0000	0 Adj R-	sq	0.6935				
C.V	•	24.4386	5	_					
	Parameter Estimates								
		Parameter	Standar	т Б	for H0:				
Variable	DF	Estimate	Errc	or Para	ameter=0	Prob >  T			
INTERCEP	1	-21570	1823.860187	4	-11.826	0.0001			
TIME	1	0.253081	0.0911830	3	2.776	0.0056			
MP	1	-1824.539417	309.6515608	6	-5.892	0.0001			
FAMLYINC	l	0.295782	0.0123131	.3	24.022	0.0001			
TEMP	1 1 1	193.357955	8.3222475	2	23.234	0.0001			
DAYS	l	274.499875	49.9306166	1	5.498	0.0001			
COMPROXY	1	6504.072456	1632.070039	0	3.985	0.0001			
D11	1	1890.103558			6.624	0.0001			
DSEAS	1	5919.128716	355.5283429	5	16.649	0.0001			

#### Model: METROPLEX SURBURBAN: LOG-LINEAR Dependent Variable: LOGCONS

#### Analysis of Variance

Source	DF	Sum of Squares	Mean Square	F Value	Prob>F
Model Error C Total	8 1443	172.54513 63.49395	21.56814 0.04400	490.170	0.0000
C TOLAL	1451	236.03907			

Root MSE	0.20976	R-square	0.7310
Dep Mean	9.56174	Adj R-sq	0.7295
C.V.	2.19380		

#### Parameter Estimates

 Variable	DF	Parameter Estimate	Standard Error	T for HO: Parameter=0	Prob >  T
 INTERCEP	1	7.144859	0.10124033	70.573	0.0000
TIME	1	0.000021755	0.00000506	4.298	0.0001
MP	1	-0.137026	0.01718839	-7.972	0.0001
 FAMLYINC	1	0.000019751	0.0000068	28.897	0.0001
TEMP	1	0.013067	0.00046196	28.286	0.0001
DAYS	1	0.016637	0.00277159	6.003	0.0001
COMPROXY	1	0.341381	0.09059428	3.768	0.0002
 D11	1	0.112936	0.01584002	7.130	0.0001
DSEAS	l	0.251443	0.01973496	12.741	0.0001

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# --Model: METROPLEX SUBURBAN: LOG-LOG Dependent Variable: LOGCONS

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	Analysis of Variance									
	_		5 <b>-</b>	Sum of		Mean				
	Source		DF	Squares	sq	luare	F Val	Lue	Prob>F	
	Model			2.79151		9894	492.7	782	0.0000	
	Error C Total			3.24756 6.03907	0.0	4383				
يومود دواوني مراجع										
		t MSE Mean	0.209 9.561		square Ij R-sq		7320 7306			
	C.V		2.189		קישיין ו	0.	, 300			
			Pa	rameter E	stimate	S				
			10		15 C L Ma CC					
			Parameter		ndard	T for I		_		
	Variable	DF	Estimate		Error	Paramet	er=0	Prob >	T	
	INTERCEP	l	-5.139531	0.583	84730	-8	.803	0.	0001	
	LOGTIME	1	0.135581		35433		.201		0014	
	LOGMP	1 1	-0.124848		25116		.611		0001	
	LOGFINC	1	0.842886		61132		.535		0001	
	LOGTEMP	1	0.734592		97762		.230		0001	
	LOGDAYS	1	0.467252		52911		.355		0001	
_	LOGPROXY	l	0.101949		70706		.503		0001	
	D11	1	0.078444		79085		.968		0001	
	DSEAS	1	0.305530	0.018	76254	16	.284	0.	0001	

METROPLEX SUBURBAN

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#### CORRELATION ANALYSIS

	6 'VAR' Variables:	TIME	MP	FAMLYINC	TEMP	DAYS	COMPROXY
			Simple Stati	stics			
98	Variable	N	Mea	n	Std Dev		Sum
	TIME	1452	856	8	1160	1244	0472
	MP	1452	1.3279	0	0.38287		1928
- 184 - 194	FAMLYINC	1452	4848	2	10777	7039	6490
	TEMP	1452	65.1711	9 :	15.16228	9	4629
	DAYS	1452	27.5482	1	2.15550	4	0000
tanan chi	COMPROXY	1452	0.0913	9	0.06618	132.6	9237

# Simple Statistics

 Variable	Minimum	Maximum
TIME	6575	10562
 MP	0.64286	2.30438
FAMLYINC	26312	87883
TEMP	32.58500	91.98500
 DAYS	19.00000	31.00000
 COMPROXY	0.01000	0.47917

- Pearson Correlation Coefficients / Prob > |R| under Ho: Rho=0 / N = 1452

	TIME	MP	FAMLYINC	TEMP	DAYS	COMPROXY
TIME	1.00000	0.30441	0.00895	0.04390	-0.03705	0.00738
	0.0	0.0001	0.7332	0.0945	0.1582	0.7788
— мр	0.30441	1.00000	-0.16430	-0.09208	-0.04319	-0.08983
	0.0001	0.0	0.0001	0.0004	0.0999	0.0006
- FAMLYINC	0.00895	-0.16430	1.00000	-0.01140	-0.06932	0.37098
	0.7332	0.0001	0.0	0.6642	0.0082	0.0001
TEMP	0.04390	-0.09208	-0.01140	1.00000	0.09785	-0.01485
	0.0945	0.0004	0.6642	0.0	0.0002	0.5717
DAYS	-0.03705	-0.04319	-0.06932	0.09785	1.00000	-0.02979
	0.1582	0.0999	0.0082	0.0002	0.0	0.2567
COMPROXY	0.00738 0.7788	-0.08983 0.0006	0.37098 0.0001	-0.01485 0.5717	-0.02979 0.2567	1.00000

### 4odel: ROLLING PLAINS REGION: LINEAR Dependent Variable: CONS

# Analysis of Variance

<b>10 1</b> 1	Source	DF	Sum Squar		Mean Square	F Value	Prob>F
	Model	10	264307057	41	2643070574.1	137.263	0.0001
	Error	517	9955096869	.0	19255506.516		
	C Total	527	363858026	10			
	Root MSE	438	88.10967	F	R-square	0.7264	
ana~1	Dep Mean	2117	2.86364	A	Ndj R-sq	0.7211	
	c.v.	2	20.72516				

#### Parameter Estimates

-	Variable	DF	Parameter Estimate	Standard Error	T for H0: Parameter=0	Prob >  T
	INTERCEP	1	-11913	6064.0404578	-1.964	0.0500
	TIME	1	-0.776597	0.20586188	-3.772	0.0002
	MP	1	-5931.982953	1269.1684474	-4.674	0.0001
	FAMLYINC	1	0.547395	0.19111070	2.864	0.0043
	TEMP	1	286.207899	16.33152561	17.525	0.0001
	DAYS	1	200.609337	97.19864266	2.064	0.0395
	COMPROXY	1	33642	16283.981798	2.066	0.0393
	D1	1	-9059.204931	2573.5042314	-3.520	0.0005
	D21	1	-7901.579490	2367.1474687	-3.338	0.0009
	D28	1	-9643	2826.3196940	-3.412	0.0007
	DSEAS	1	5345.606923	635.77889602	8.408	0.0001

# Model: ROLLING PLAINS REGION: LOG-LINEAR Dependent Variable: LOGCONS

			Analysi	ls of Var	iance		
se the	Source	DF	Sum Squar		Mean Square	F Value	Prob>F
<b>10</b>	Model Error C Total	10 517 527	61.401 23.221 84.623	64	6.14019 0.04492	136.704	0.0001
jagan sen	Root MSE Dep Mean C.V.	9	0.21193 9.88248 2.14454	R-squa Adj R-		0.7256 0.7203	

#### Parameter Estimates

 Variable	DF	Parameter Estimate	Standard Error	T for HO: Parameter=0	Prob >  T
INTERCEP	1	8.015805	0.29287736	27.369	0.0001
TIME	1	-0.000028619	0.00000994	-2.878	0.0042
 MP	1	-0.299427	0.06129753	-4.885	0.0001
 FAMLYINC	1	0.000030872	0.00000923	3.345	0.0009
TEMP	1	0.015248	0.00078877	19.332	0.0001
DAYS	1	0.009884	0.00469444	2.105	0.0357
 COMPROXY	1	1.354971	0.78647391	1.723	0.0855
D1	1	-0.487448	0.12429355	-3.922	0.0001
D21	1	-0.418419	0.11432706	-3.660	0.0003
 D28	1	-0.521871	0.13650388	-3.823	0.0001
DSEAS	1	0.172291	0.03070646	5.611	0.0001

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# lodel: ROLLING PLAINS REGION: LOG-LOG Dependent Variable: LOGCONS

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	- Analysis of Variance										
	Source		DF	Sum of Squares	Sc	Mean Juare F	Value	Prob>F			
	Model Error C Total		517 23	0.72633 3.89725 4.62358		)7263 1 )4622	31.377	0.0001			
		ht MS Mea		48 Ad	-square lj R-sq	0.717 0.712					
—	Parameter Estimates										
	Variable	DF	Parameter Estimate	Sta	andard Error	T for H0: Parameter=	0 Prob	) >  T			
	INTERCEP TIME LOGMP LOGFINC	1 1 1 1	-6.442416 -0.000029889 -0.275934 1.198870	0.000 0.053	235629 00988 868670 294393	-1.66 -3.02 -5.14 3.30	5 0	0.0968 0.0026 0.0001 0.0010			
	LOGTEMP LOGDAYS LOGPROXY D1	1 1 1 1	0.871546 0.233351 0.122509 -0.500443	0.046 0.120 0.061	96277 88692 21544 09892	18.55 1.93 2.00 -4.09	8 0 1	0.0001 0.0541 0.0459 0.0001			
	D21 D28 DSEAS	1 1 1	-0.427581 -0.518969 0.224957	0.115 0.134	29756 25101 81238	-3.70 -3.86 7.54	8 6	0.0002 0.0001 0.0001			

ROLLING PLAINS REGION

0.0001

576

0.0001

528

#### CORRELATION ANALYSIS 6 'VAR' Variables: TIME MP FAMLYINC TEMP DAYS COMPROXY Simple Statistics Variable Ν Std Dev Mean Sum TIME 576 8385 1266 4829880 0.79049 MP 528 0.19714 417.37956 FAMLYINC 576 40442 4822 23294502 TEMP 576 66.27068 15.08794 38172 DAYS 576 27.96528 2.01529 16108 56.88847 COMPROXY 576 0.09876 0.03096 Simple Statistics Variable Minimum Maximum TIME 6210 10562 MP 0.52321 1.36442 30353 FAMLYINC 45445 TEMP 30.08000 91,90500 DAYS 13.00000 31.00000 COMPROXY 0.03000 0.16000 Pearson Correlation Coefficients / Prob > |R| under Ho: Rho=0 / Number of Observations TIME MP FAMLYINC TEMP DAYS COMPROXY 1.00000 0.10292 0.19152 -0.04857- TIME 0.01078 0.38428 0.0 0.0180 0.0001 0.7962 0.2445 0.0001 576 576 576 528 576 576 0.10292 1.00000 ¯ MΡ 0.58620 -0.18135-0.11763-0.384120.0180 0.0001 0.0 0.0001 0.0068 0.0001 528 528 528 528 528 528 -0.13207 FAMLYINC 0.19152 0.58620 1.00000 -0.25169 -0.32234 0.0001 0.0001 0.0 0.0001 0.0015 0.0001 576 528 576 576 576 576 TEMP 0.01078 1.00000 -0.18135-0.25169-0.03190 0.14894 0.7962 0.0001 0.0001 0.0 0.4448 0.0003 576 528 576 576 576 576 -0.04857-0.11763 -0.13207 1.00000 DAYS -0.03190 -0.02403 0.2445 0.0068 0.0015 0.4448 0.0 0.5649 576 528 576 576 576 576 COMPROXY 0.38428 -0.38412-0.32234 0.14894 -0.024031.00000

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576

0.0001

0.0003

576

0.5649

576

0.0

576

# Model: SOUTHEAST REGION: LINEAR pependent Variable: CONS

Analysis	of	Variance
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<b></b>	Source	DF	Sum of Squares		F Value	Prob>F
	Model Error			1643659497.2 5270523.1029	311.859	0.0000
ANNO 10	C Total		798506331	5270525.1029		
	Root MSE	2295.7	-	R-square	0.7781	
and a	Dep Mean C.V.	15434.2 14.8	17446	Adj R-sq	0.7756	

# Parameter Estimates

Variable	DF	Parameter Estimate	Standard Error	T for HO: Parameter=0	Prob >  T
 INTERCEP	1	-1996.386490	1154.4456417	-1.729	0.0840
TIME	1	-0.181954	0.06163899	-2.952	0.0032
MP	1	-1395.244115	236.28005589	-5.905	0.0001
 FAMLYINC	1	0.237526	0.01331346	17.841	0.0001
TEMP	1	86.421350	6.30083100	13.716	0.0001
DAYS	l	96.150638	25.70203944	3.741	0.0002
 COMPROXY	1	4990.709171	1262.0879326	3.954	0.0001
D8	1	6408.858508	262.22546199	24.440	0.0001
GAL	1	4732.750454	276.68752336	17.105	0.0001
LEA	1	-1648.659662	339.59336569	-4.855	0.0001
 TEX	1	-1009.821740	259.24902778	-3.895	0.0001
ALV	1	-1409.286176	253.19825131	-5.566	0.0001
ANG	1	-2687.339961	252.73317654	-10.633	0.0001
 BRA	1	-4057.631951	297.05337000	-13.660	0.0001
FRE	1	5035.064919	313.38021464	16.067	0.0001
LAK	l	-2423.920637	252.37558931	-9.604	0.0001
 DSEAS	1	1349.493424	195.41129353	6.906	0.0001

# Model: SOUTHEAST REGION: LOG-LINEAR Dependent Variable: LOGCONS

• **1				Analysis	of Varian	nce			
	Source		DF	Sum of Squares		Mean quare	F Val	lue	Prob>F
-del tran	Model Error C Total		16 1423 1439	115.17673 27.21116 142.38789	0.0	19855 )1912	376.4	146	0.0000
-arr-a,		t MSI Mear	n 9.5		R-square Adj R-sq		.8089 .8067		
				Parameter	Estimate	es			
<b>1000</b> , 700,			Paramet	er S	tandard	T for	H0:		
	Variable	DF	Estima		Error	Paramet		Prob >	т
-	INTERCEP TIME MP	1 1 1	8.4932 -0.0000134 -0.0696	42 0.0	6953729 0000371 1423218	-3	2.139 3.620 4.895	0.00 0.00 0.00	03
	FAMLYINC TEMP DAYS	1 1 1	0.0000153 0.0057 0.0061	805 0.0 717 0.0	0000080 0037953 0154815	19 15	0.085 5.064 8.998	0.00	01 01
	COMPROXY D8 GAL	1 1 1	0.2373 0.3408 0.2760	.0 080	7602105 1579498 1666609	21	3.122 .582 5.565	0.00 0.00 0.00	18 01
	LEA TEX ALV	1 1 1	-0.1383 -0.0608 -0.0940	0.0	2045519 1561570 1525123	- 3	5.762 3.894 5.169	0.00 0.00 0.00	01 01
	ANG BRA FRE	1 1 1	-0.1907 -0.3976 0.2868	05 0.0 27 0.0	1522322 1789282 1887625	-12 -22	2.527	0.00 0.00 0.00	01 01
_	LAK DSEAS	1 1	-0.1788 0.0688	45 0.0	1520168 1177047	-11	.765	0.00	01

# Model: SOUTHEAST REGION: LOG-LOG Dependent Variable: LOGCONS

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# Analysis of Variance

	Source		DF		m of ares		Mean uare	F Va	lue	Prob>F
r-	Model Error C Total		16 1423 1439	114.9 27.4 142.3	2351		8527 1927	372.	842	0.0000
~		t MSE Mean •	9.	13882 59558 44673		square j R-sq		).8074 ).8052		
				Param	eter E	stimate	S			
			Parame	tor	Sta	ndard	T for	- uo.		
	Variable	DF	Estim			Error	Parame		Prob >	T
	INTERCEP	1	1.389	987	0.617	90246		2.250	0.	0246
	LOGTIME	1	-0.152	247	0.0309	95027	-	4.919	0.	0001
	LOGMP	1	-0.087	162	0.0200	05363	-	4.346	Ο.	0001
	LOGFINC	1	0.718	286	0.0409	94729	1	7.542	0.	0001
	LOGTEMP	1	0.354	680	0.0243	38446	1	4.545	Ο.	0001
	LOGDAYS	1	0.142	706	0.0402	28980		3.542	0.	0004
	LOGPROXY	1	0.037	680	0.0096	56490		3.899	0.	0001
	D8	1	0.330	731	0.0158	34576	2	0.872	0.	0001
	GAL	1	0.280	520	0.0169	95594	1	6.544	0.	0001
	LEA	1	-0.145	589	0.0193	79059	-	7.356	Ο.	0001
	TEX	1	-0.044	898	0.0174	10176	-	2.580	Ο.	0100
	ALV	1	-0.086	214	0.0153	32686	-	5.625	0.	0001
	ANG	1	-0.195	783	0.0153	36812	-1	2.740	0.	0001
	BRA	1	-0.380	608	0.0183	30996	-2	0.787	0.	0001
	FRE	1	0.297	972	0.0173	L4965	1	7.375	Ο.	0001
	LAK	1	-0.179	347	0.0153	33369	-1	1.696	0.	0001
	DSEAS	1	0.080	547	0.0114	19926		7.005	0.	0001

SOUTHEAST REGION

#### CORRELATION ANALYSIS

	6 'VAR'	Variables:	TIME M	ip Fi	AMLYINC TEMP	DAYS	COMPROXY
			Simp	le Statis	tics		
	Variab]	le	N	Mean	Std 1	Dev	Sum
-	TIME MP FAMLYIN	ĩC	1440 1440 1440	8565 1.32836 47817	0.45	820 993 6	12333272 1913 58857159
	TEMP DAYS COMPROX	۲Y	1440 1440 1440	68.78029 26.77361 0.10522	11.644 2.419 0.074	955	99044 38554 51.52273
	Variabl	-	Statistics	Maximum			
	TIME MP FAMLYIN		6575 .65268 30266	10562 3.28659 65054			
	TEMP DAYS COMPROX	43 18	.13000 .00000 .02020	87.87000 31.00000 0.40404			
_	Pearson Co	rrelation	Coefficients	/ Prob >	R  under Ho	: Rho=0 /	N = 1440
		TIME	MP	FAMLYINC	TEMP	DAYS	COMPROXY
	TIME	1.00000 0.0	0.29621 0.0001	-0.07862 0.0028	0.06388 0.0153	0.05639 0.0324	0.17312 0.0001
	MP	0.29621 0.0001	1.00000 0.0	0.16675 0.0001	0.00753 0.7754	0.01296 0.6232	0.28966 0.0001
—	FAMLYINC	-0.07862 0.0028	0.16675 0.0001	1.00000 0.0	-0.01030 0.6960	-0.01305 0.6206	0.48177 0.0001
-	TEMP	0.06388 0.0153	0.00753 0.7754	-0.01030 0.6960	1.00000 0.0	-0.00211 0.9362	0.01014 0.7005
	DAYS	0.05639 0.0324	0.01296 0.6232	-0.01305 0.6206	-0.00211 0.9362	1.00000 0.0	-0.01926 0.4653
-	COMPROXY	0.17312 0.0001	0.28966 0.0001	0.48177 0.0001	0.01014 0.7005	-0.01926 0.4653	1.00000 0.0

# Model: VALLEY REGION: LINEAR Dependent Variable: CONS

# Analysis of Variance

Source	DF	Sum of Squares		F Value	Prob>F
Model Error C Total	11 912 923		5 2122806565.9 5 24939719.053 2	85.118	0.0001
Root MSE Dep Mean C.V.	2295	3.96827 4.61147 1.75584	R-square Adj R-sq	0.5066 0.5006	

#### Parameter Estimates

 Variable	DF	Parameter Estimate	Standard Error	T for H0: Parameter=0	Prob >  T
 INTERCEP TIME	ן ו	5505.802596 -0.829722	3664.6024603 0.16237484	1.502 ~5.110	0.1333 0.0001
MP	ī	-8656.641869	1264.4735182	-6.846	0.0001
 FAMLYINC	1	0.220667	0.04318008	5.110	0.0001
TEMP	1	203.516656	19.09267294	10.659	0.0001
DAYS	1	409.280531	86.11968404	4.752	0.0001
COMPROXY	1	24930	4142.5910635	6.018	0.0001
 MCA	1	3543.464684	522.40251895	6.783	0.0001
EDI	1	1856.729183	623.37364733	2.979	0.0030
PHA	1	905.216465	691.43213718	1.309	0.1908
 HAR	l	-6967.380343	665.20517369	-10.474	0.0001
DSEAS	1	3145.192855	522.37571337	6.021	0.0001

#### Model: VALLEY REGION: LOG-LINEAR )ependent Variable: LOGCONS

0.000010704

0.009106

0.017671

1.042447

0.180689

0.112436

0.072082

0.115354

-0.299632

FAMLYINC

COMPROXY

TEMP

DAYS

MCA

EDI

PHA

HAR

DSEAS

1

1

1

1

1

1

1

1 1

Analysis of Variand
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	Source		DF	Sum of Squares		Mean Juare	F Val	lue	Prob>F
4. mark	Model Error C Total		912 4	15.55071 10.69554 36.24625		4097 4462	92.8	301	0.0001
der, song		t MSE Mear •		500 2	R-square Adj R-sq		.5281 .5225		
			Pa	arameter	Estimate	S			
4	Variable	DF	Parameter Estimate		tandard Error	T for Paramet		Prob >	T
	INTERCEP TIME MP	1 1 1	9.211863 -0.000036579 -0.411456	0.00	5500906 0000687 5348598	-5	9.428 5.326 7.693	0.0 0.0 0.0	

0.0000183

0.00080760

0.00364278

0.17522751

0.02209711

0.02636809

0.02924690

0.02813752

0.02209598

5.861

4.851

5.949

8.177

4.264

2.465

5.221

-10.649

11.276

0.0001

0.0001

0.0001

0.0001

0.0001

0.0001

0.0139

0.0001

0.0001

### Model: VALLEY REGION: LOG-LOG pependent Variable: LOGCONS

<b></b>			A	nalysis	of Varia	nce				
<b>.</b>	Source		DF	Sum of Squares		Mean quare	F Va	alue	Prob>F	
	Model Error C Total		11 912 923	45.03913 41.20712 86.24625		09447 04518	90	.619	0.0001	
		MSE Mean	0.21 9.99 2.12	500 .	R-square Adj R-sq		0.5222 0.5165			
	Parameter Estimates									
	Variable	DF	Paramete Estimat		tandard Error		or HO: neter=0	Prob >	T	
	INTERCEP LOGTIME	1	8.96125	0 0.0	3180172 5561928		12.245	0.	0001 0001	
	LOGMP LOGFINC LOGTEMP	1 1 1 1	-0.38912 0.05528 0.61765	1 0.0 3 0.0	4188872 2653176 5604660		-9.289 2.084 11.020	0.	0001 0375 0001	
	LOGDAYS LOGPROXY MCA	1 1 1	0.41496 0.14190 0.15990	4 0.0 1 0.0	9606856 1713603 2238454		4.319 8.281 7.143	0.	0001 0001 0001	
	EDI PHA HAR DSEAS	1 1 1 1	0.16206 0.14260 -0.39777 0.12658	4 0.02 3 0.02	2483033 2746259 2691119 2175719	-	6.527 5.193 -14.781 5.818	0.	0001 0001 0001 0001	

VALLEY REGION

#### CORRELATION ANALYSIS

	6 'VAR'	Variables:	TIME	MP FA	AMLYINC TEMP	DAYS	COMPROXY
			Sir	nple Statis	tics		
•.	Variabl	e	N	Mean	Std	Dev	Sum
₩ t st	TIME MP FAMLYIN TEMP DAYS COMPROX		924 924 924 924 924 924 924	8568 0.81986 31167 73.57111 28.04870 0.11519	0.20	528 75 216 2 188 581	7916664 7.55452 8798617 67980 25917 6.43547
			Statistics				
100 ma	Variabl	e M	inimum	Maximum			
	TIME MP FAMLYIN	С	6575 46012 2993	10562 1.29199 48955			
-	TEMP DAYS COMPROX	16	.12500 .00000 .01010	89.61000 31.00000 0.36842			
P	earson Co	rrelation (	Coefficient	ts / Prob >	R  under H	o: Rho=0 / 1	N = 924
_		TIME	МР	FAMLYINC	TEMP	DAYS	COMPROXY
T	IME	1.00000 0.0	0.28738 0.0001	-0.19077 0.0001	0.05093 0.1219	0.03006 0.3614	0.03664 0.2659
M	P	0.28738 0.0001	1.00000 0.0	0.03932 0.2324	0.00483 0.8833	-0.05041 0.1257	-0.01003 0.7608
— F/	AMLYINC	-0.19077 0.0001	0.03932 0.2324	1.00000 0.0	0.00101 0.9756	-0.00437 0.8944	0.30401 0.0001
— TI	EMP	0.05093 0.1219	0.00483 0.8833	0.00101 0.9756	1.00000 0.0	-0.12053 0.0002	0.00581 0.8600
D/	AYS	0.03006 0.3614	-0.05041 0.1257	-0.00437 0.8944	-0.12053 0.0002	1.00000 0.0	-0.00582 0.8598
_ C	OMPROXY	0.03664 0.2659	-0.01003 0.7608	0.30401 0.0001	0.00581 0.8600	-0.00582 0.8598	1.00000 0.0

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4odel: WEST REGION: LINEAR
Dependent Variable: CONS

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# Analysis of Variance

<b>199</b> * *	Source	DF	Sum o Square		F Value	Prob>F
	Model Error C Total	6 653 659		0 5222041966.7 9 17567575.145 0	297.255	0.0001
	Root MSE Dep Mean C.V.	21941	36913 36818 10259	R-square Adj R-sq	0.7320 0.7295	
			Paramete	r Estimates		

 Variable	DF	Parameter Estimate	Standard Error	T for H0: Parameter=0	Prob >  T
INTERCEP	1	-36654	3412.3052204	-10.742	0.0001
MP	1	-1929.111210	717.35804521	-2.689	0.0073
 FAMLYINC	1	0.122150	0.02604647	4.690	0.0001
TEMP	1	382.317129	20.79378033	18.386	0.0001
DAYS	1	1021.504243	104.76109354	9.751	0.0001
 COMPROXY	1	8231.467259	2180.7883385	3.775	0.0002
DSEAS	1	3141.149196	647.23688346	4.853	0.0001

# Aodel: WEST REGION: LOG-LINEAR Dependent Variable: LOGCONS

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# Analysis of Variance

<b>~</b> .	Source	DF	Sum of Squares		F Value	Prob>F
∎∎ sða	Model Error C Total	6 653 659	63.99562 20.74412 84.73973	0.03177	335.751	0.0001
	Root MSE Dep Mean C.V.	9		R-square Adj R-sq	0.7552 0.7530	

# Parameter Estimates

,5049 Ma	Variable	DF	Parameter Estimate	Standard Error	T for HO: Parameter=0	Prob >  T
	INTERCEP	l	7.381892	0.14510516	50.873	0.0001
	MP	1	-0.063118	0.03050500	-2.069	0.0389
	FAMLYINC	1	0.000005209	0.00000111	4.703	0.0001
	TEMP	1	0.018515	0.00088424	20.938	0.0001
	DAYS	1	0.040344	0.00445487	9.056	0.0001
	COMPROXY	1	0.384444	0.09273603	4.146	0.0001
	DSEAS	1	0.097948	0.02752316	3.559	0.0004

4odel: WEST REGION: LOG-LOG Dependent Variable: LOGCONS

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# Analysis of Variance

<b>-</b> -	Source	DF	Sum ( Squar)		Mean quare	F Val	ue	Prob>F
<b>-</b> • •	Model Error C Total	6 653 659	61.816 22.923 84.739	05 0.	30278 03510	293.4	91	0.0001
<b>#</b> * **	Root MSE Dep Mean C.V.	9.	18736 93168 88650	R-square Adj R-sq		0.7295 0.7270		
			Paramete	er Estimat	es			
	Variable DF	Parame Estim		Standard Error		or HO: neter=0	Prob >	T

	variable	DF.	Estimate	Error	Parameter=0	
	INTERCEP	1	-0.857688	0.76959397	-1.114	0.2655
	LOGMP	1	-0.099418	0.03133803	-3.172	0.0016
-	LOGFINC	1	0.305497	0.05608852	5.447	0.0001
	LOGTEMP	1	0.866224	0.04760267	18.197	0.0001
	LOGDAYS	1	1.187610	0.12995848	9.138	0.0001
	LOGPROXY	1	0.040590	0.01467332	2.766	0.0058
	DSEAS	1	0.205719	0.02587832	7.949	0.0001

WEST REGION

#### CORRELATION ANALYSIS

. 6	5 'VAR'	Variables:	MP	FAMLYINC T	EMP DAY	S COMPRO	OXY DSEAS				
Simple Statistics											
1	Variab]	e	N	Mean	std	Dev	Sum				
y (196	MP FAMLYIN TEMP DAYS COMPROX DSEAS		660 720 720 720 720 720	0.97979 46401 61.31661 28.60139 0.11817 0.41667	15.0 1.6 0.0	8231 7809 7057 7649 8	46.66131 33408653 44148 20593 35.08445 00.00000				
<b>8</b> . 100		Simple	Statistics								
	Variabl	.e M	inimum	Maximum							
		IC 24 22 IY 0		1.83861 61683 87.14500 31.00000 0.41000 1.00000	Prob >  R  1	under Ho: Rf	10=0				
		MP	FAMLYINC	ТЕМР	DAYS	COMPROXY	DSEAS				
MP		1.00000 0.0 660	0.63148 0.0001 660	-0.00873 0.8229 660	-0.11917 0.0022 660	0.19112 0.0001 660	-0.04058 0.2979 660				
- Fam	LYINC	0.63148 0.0001 660	1.00000 0.0 720	0.07132 0.0558 720	-0.12847 0.0005 720	0.16096 0.0001 720	0.00000 1.0000 720				
- Tem	P	-0.00873 0.8229 660	0.07132 0.0558 720	1.00000 0.0 720	-0.25119 0.0001 720	-0.06139 0.0998 720	0.84892 0.0001 720				
DAY.	S	-0.11917 0.0022 660	-0.12847 0.0005 720	-0.25119 0.0001 720	1.00000 0.0 720	-0.11694 0.0017 720	-0.32640 0.0001 720				
СОМ —	PROXY	0.19112 0.0001 660	0.16096 0.0001 720	-0.06139 0.0998 720	-0.11694 0.0017 720	1.00000 0.0 720	0.00000 1.0000 720				
DSE	AS	-0.04058 0.2979 660	0.00000 1.0000 720	0.84892 0.0001 720	-0.32640 0.0001 720	0.00000 1.0000 720	1.00000 0.0 720				

# APPENDIX B

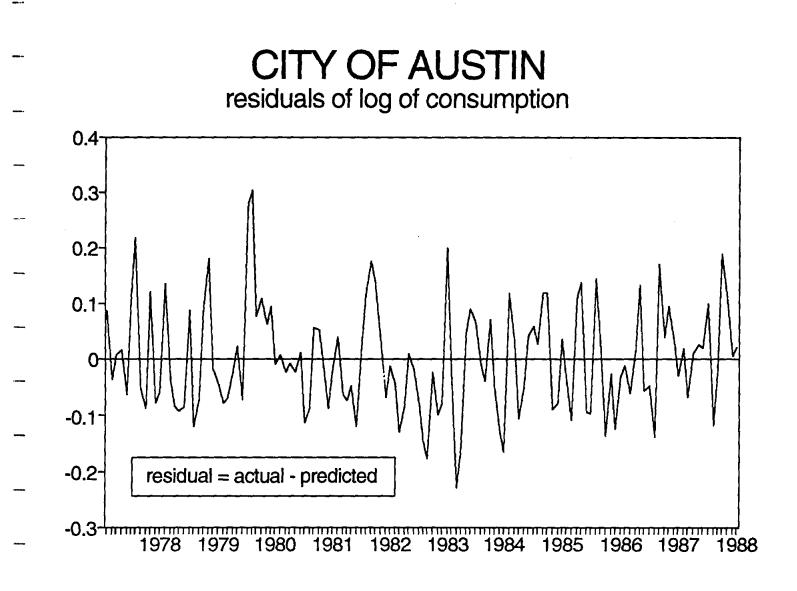
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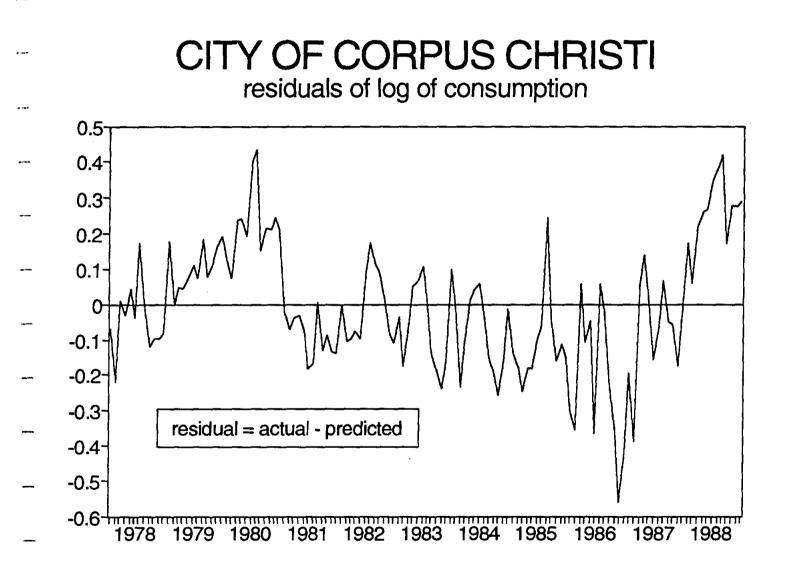
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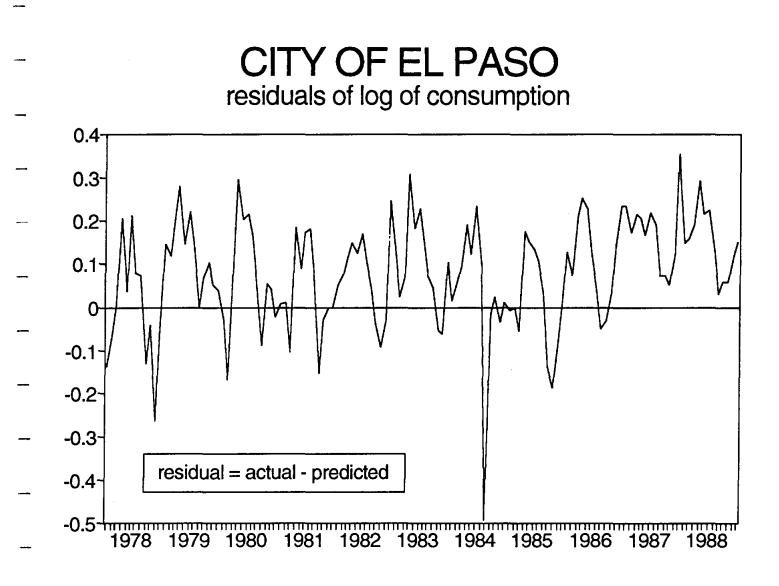
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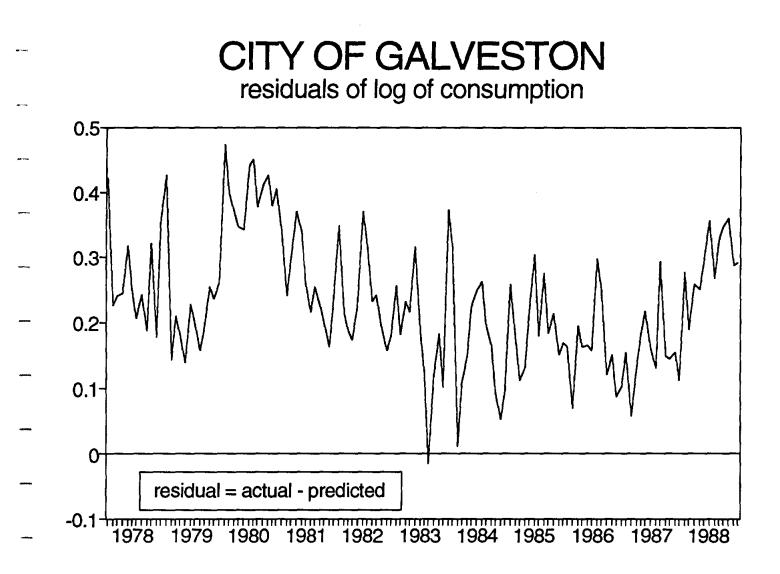
Residuals Comparisons for Selected Cities

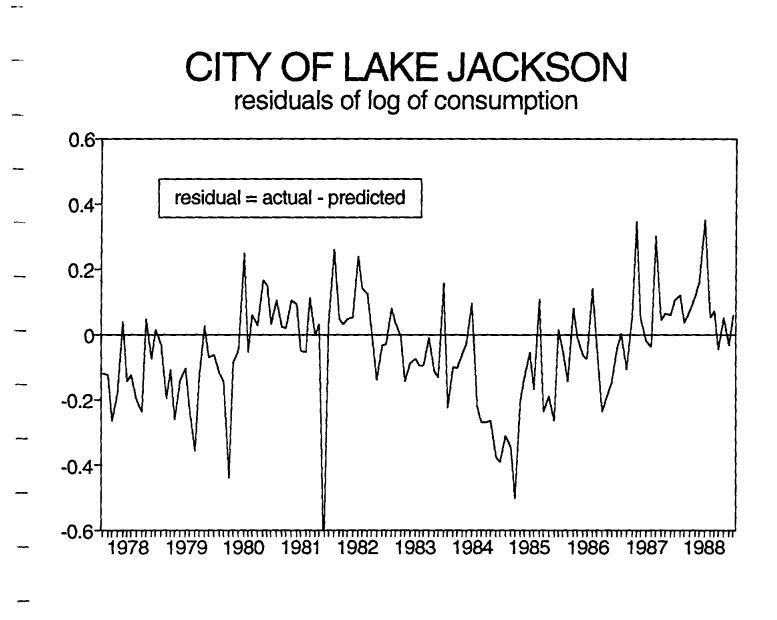


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