



# Evaluation of controllable and uncontrollable drivers on residential water demand: An application to the Las Vegas area

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NWRA, 2014 Annual Conference  
February 6<sup>th</sup>, 2014

# ACKNOWLEDGMENTS

This work has been sponsored by Nevada EPSCoR.

We are grateful to individuals at Southern Nevada Water Authority for providing information and feedback.

# OUTLINE

1. Factors affecting residential water demand
  - i. Uncontrollable by utility: exogenous factors
  - ii. Controllable by utility: factors that alter behavior of customers through conservation policies
2. Model formulation
3. Descriptions of variables
3. Data sources
4. Results and Projections
5. Conclusion

# WATER DEMAND DRIVERS AFFECT CUSTOMER DECISIONS AND BEHAVIOR

## Residential Water Demand



- Avg Temperature
- Avg Wind Speed
- Precipitation Days

- Seasonality
- Economic trends

- Household Income
- Household Size

Exogenous, uncontrollable by policy; but can affect policy results

- Marginal Price

Price Policies

- Turf Restriction Regulation

Non-Price Policies

- Turf Area
- Treed Area
- Non Irrigated Area
- Swimming pools

Controllable: These drivers influence water through policies

# We explicitly model the effects of these drivers on customers' behavior

- Use observed data to predict monthly water use as a function of variables that represent controllable and uncontrollable drivers (next slides)
- Observations over time to account for seasonality and economic trends
- Observations at customer level to account for household differences
- Uncontrollable drivers: account for how exogenous effects influence how customers respond to controllable drivers (policies).
- A random effects model accounts for heterogeneity of households and time invariant variables
- Use estimated model to:
  - predict how customers will respond to multiple policy tools, as represented by systematic changes in controllable drivers
  - predict how responses to policies vary with systemic changes in uncontrollable drivers.

# Random Effect Model

For a given household  $i$ , water demand in month  $t$  can be expressed as

$$y_{it} = \hat{p}_{it}\beta_p + \hat{d}_{it}\beta_d + x'_i\beta_x + c'_t\beta_c + m'_t\beta_m + \varepsilon_{it} \quad (1)$$

**where**  $\varepsilon_{it} = \mu_i + e_{it}$ ,

**and**

$$p_{it} = b'_i\beta_{b_{it}} + x'_i\beta_x + c'_t\beta_c + m'_t\beta_m + \varepsilon_{it1} \quad (2)$$

$$d_{it} = b'_i\beta_{b_{it}} + x'_i\beta_x + c'_t\beta_c + m'_t\beta_m + \varepsilon_{it2} \quad (3)$$

## Variables:

$y_{it}$  - household monthly water use in 1000s of gallons

$p_{it}$  - estimated marginal price associated with household  $i$  and month  $t$

$d_{it}$  - estimated difference variable associated with household  $i$  and month  $t$

$x_i$  - is a  $k$  by  $1$  vector of household characteristics including landscape characteristics

$c_t$  - is a  $l$  by  $1$  vector of climate indicators

$m_t$  - is a  $n$  by  $1$  vector of period indicator variables corresponding to time period  $t$ ,

$b_i$  is a  $k$  by  $1$  vector of water bill characteristics

$\mu_i$  - a household-specific random error

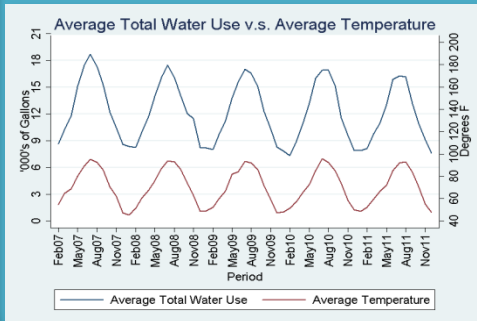
$e_{it}$  - regression random error

$\beta$  -terms are coefficients corresponding to their associated regressors

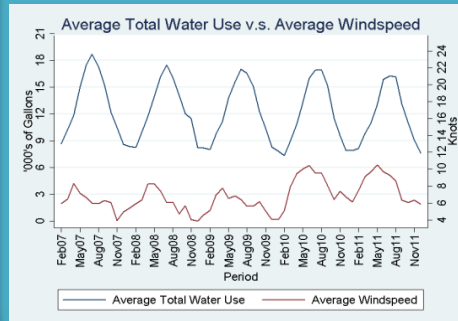
# VARIABLES TO REPRESENT UNCONTROLLABLE (EXOGENOUS) DRIVERS

Climate Variables account for weather parameters as perceived and acted on by consumers

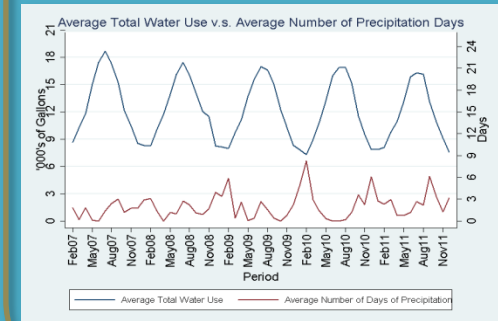
## Average Daily Temperature (F)



## Average Daily Wind Speed (Kn)



## Number of Precipitation days



# VARIABLES TO REPRESENT UNCONTROLLABLE (EXOGENOUS) DRIVERS

Household Variables:  
account for unique features of households

59 Period Variables

account for  
seasonality and  
economic trends  
(dummy variables)

Income  
(1,000 \$ dollars)

Proxy:  
Appraised House  
Value times 0.025  
(Nieswiadomy and  
Molina, 1988)

Billing  
days in  
month  
(# of days)  
Varies  
across  
sample

Household  
Size

Proxy:  
Number of  
bedrooms





# VARIABLES TO REPRESENT CONTROLLABLE DRIVERS: PRICE POLICIES

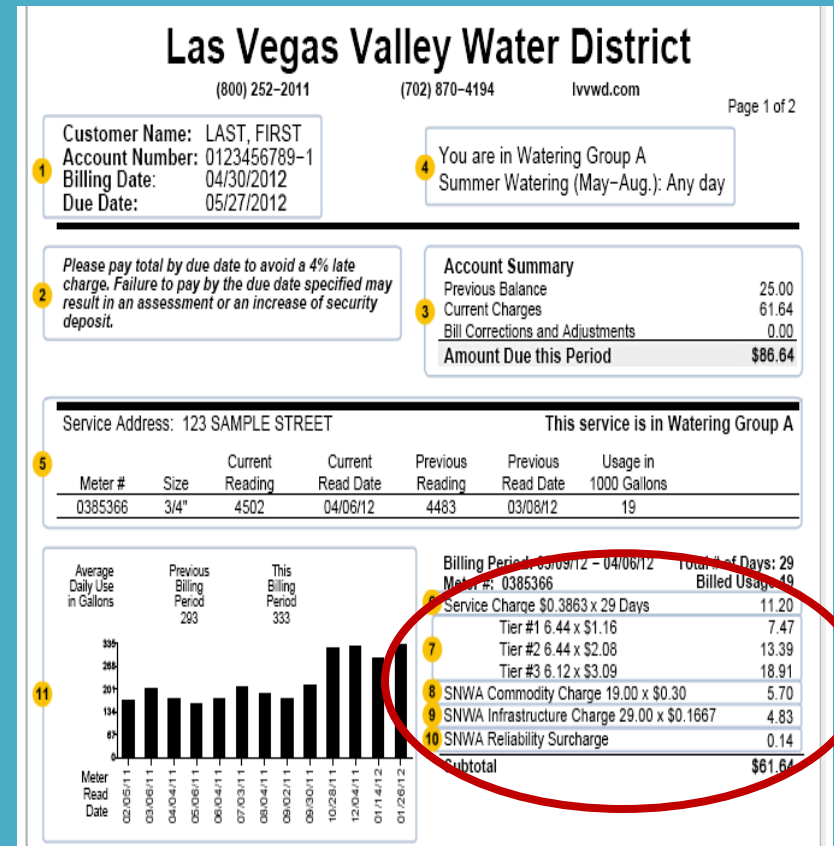
Price Variables account for block-rate structure as well as household-specific features

## Marginal price, \$

Price per unit of water at the maximum block used for each billing period per household

## Difference, \$

The difference between actual bill and bill at the marginal price; must be used with marginal price



# VARIABLES TO REPRESENT CONTROLLABLE DRIVERS: NON-PRICE POLICIES

Involuntary conservation policy:

Turf Restriction Regulation for new houses in 2003 (dummy variable)

Voluntary conservation policy:

Landscape variables represent expected effects from conservation policies that target landscaping. (Voluntary conservation program data is not used directly because of an endogeneity problem - this is future work)

Landscape Variables –  
account for unique features of  
households (from Judy Brandt's GIS  
imaging, SNWA)

Turf  
sqft

Trees & shrubs  
sqft

Non-  
Irrigated area,  
sqft  
  
Calculated as  
difference between  
yard size and sum  
of turf and treed  
areas

Swimming  
Pool,  
  
(dummy  
variable)

# SHAPING DATASET:

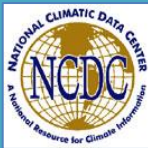
DATA WERE OBTAINED FROM FIVE NATIONAL AND REGIONAL AGENCIES .  
THIS UNIQUE, DISAGGREGATED DATASET INCLUDES ABOUT 40% OF SINGLE-FAMILY  
HOUSEHOLDS OF LAS VEGAS WITH UNINTERRUPTED WATER HISTORY FROM 2007  
TO 2011



**Southern Nevada  
Water Authority  
(SNWA)**



**LV Valley Water  
District (LVVWD)**



**National Climatic  
Data Center  
(NCDC)**



**Bureau of Labor  
Statistics**



**Clark County  
Assessor**



**Study dataset:  
62.237  
households with  
3,671,983  
monthly  
observations**

# Results

As expected:

- ✓ Difference: opposite in sign to income, thus negative (Nordin 1976)
- ✓ Marginal price: negative (law of demand)
- ✓ Water Demand Increases (+) with Bill Days, Household Size, Income, Temperature, Wind Speed, Turf, Trees, and Pool presence
- ✓ Water Demand Decreases (-) with Precipitation, Non-Irrigated area, and Turf restriction regulation implementation

Therefore, for a policy that targets replacing turf with trees :

- An estimated 55% water savings over entire sample of residential customers

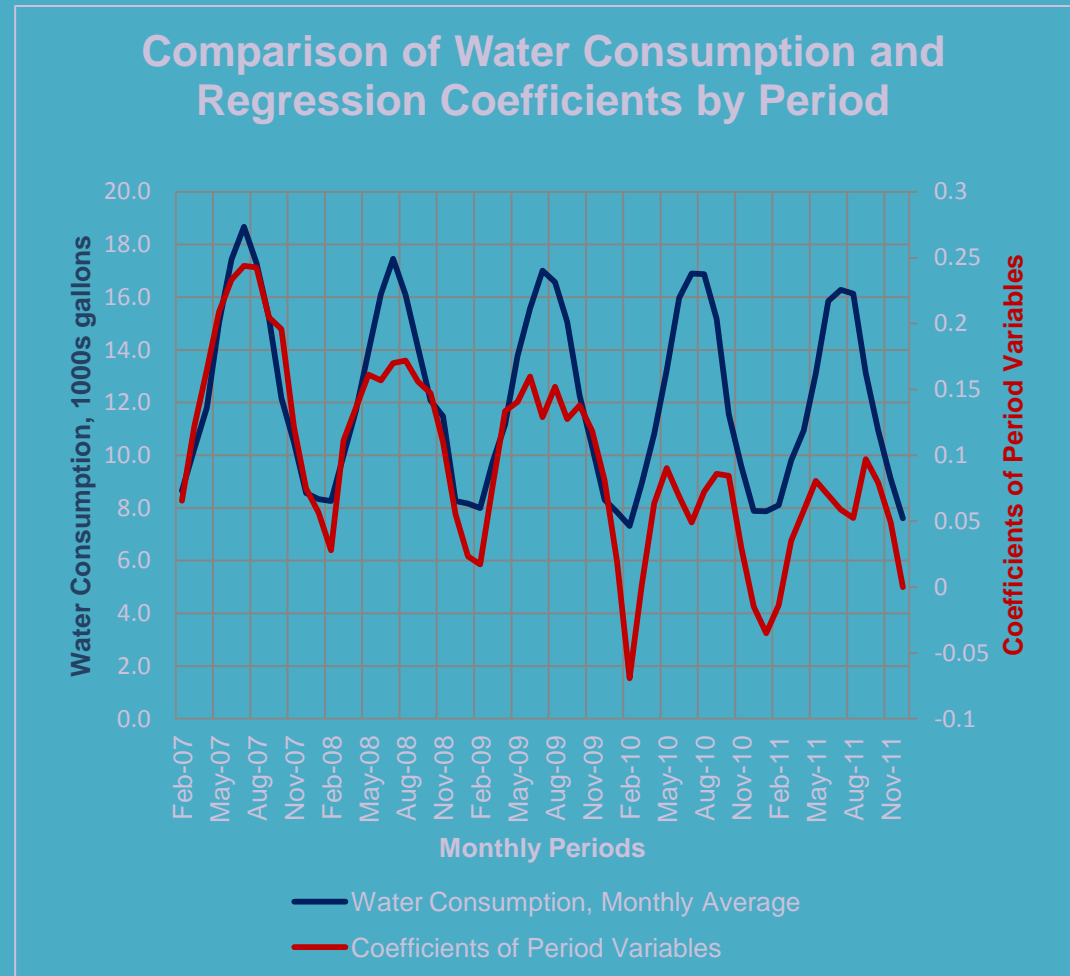
Variables	Coefficient	Std. Error
Difference	<b>-0.0203</b>	0.00148** *
Marginal price	<b>-0.1356</b>	0.01221** *
Days	<b>0.0274</b>	0.00054** *
Household size	<b>0.0869</b>	0.00188** *
Income	<b>0.00004</b>	1 E-06***
Avg, Temperature	<b>0.0090</b>	0.00018** *
Days of Precipitation	<b>-0.0082</b>	0.00018** *
Avg, Wind speed	<b>0.0153</b>	0.00041** *
Size of Turf	<b>0.00028</b>	6 E-06***
Size of Trees	<b>0.00012</b>	2 E-06***
Non-Irrigated Area	<b>0.00002</b>	6 E-06***
Turf Restriction Policy	<b>-0.0906</b>	0.00370 *
Constant	<b>0.1240</b>	0.04702**

\*\* Significance level of 5%  
 \*\*\* Significance level of 1%  
 R-sq overall = 0.4777

# Time Trends: Period Dummy Variables

## Estimated Coefficients:

- ✓ Are in accordance with seasonal changes (summer ups and winter downs)
- ✓ Clearly reflect a recession effect (decrease since 2008)



# ELASTICITIES

Elasticity measures the change in consumer water demanded as a result of a change in an explanatory variable (price, income, temperature ... )

For example:

$$\text{Price elasticity} = \text{slope} \times \frac{\text{price}}{\text{quantity of water}}$$

Important:

Price elasticity is not constant – it increases with increasing price!

Why elasticity is important?

Because elasticity is:

- ✓ Unit free
- ✓ Easily interpreted
- ✓ Comparable across studies
- ✓ Effective for exploring different scenarios

# RESULTS – ELASTICITY

## (FOR SOME REGRESSORS OF INTEREST)

- ✓ Water demand is inelastic for all presented regressors, but is responsive to change
- ✓ Demand is most responsive to temperature among non-controllable variables
- ✓ Demand is most responsive to price among controllable variables
- ✓ Price elasticity is similar to findings in other U.S. studies, confirming similar consumer behavior toward water use

Variables	Elasticity for mean of factor	Elasticity for median of factor
Marginal price	-0.315	-0.285
Family size	0.295	0.261
Size of Turf (skewed right)	0.057	0.006
Size of Trees	0.168	0.137
Income	0.145	0.134
Avg, Temperature	0.639	0.622
Days of Precipitation	-0.017	-0.016
Avg, Wind speed	0.104	0.098

\* Increase of price by 10% leads to decrease in water demand by 3.15%

# EXAMPLE OF POLICY SCENARIOS: WATER DEMAND PROJECTIONS USING ELASTICITIES

Using elasticities, it is possible to create alternative policy scenarios

Analysis of Elasticities of Manageable Variables: Price and Turf Size								
			1st Portfolio (20% of Price Increase and 30% of Turf Decrease)			2nd portfolio (30% Price increase and 20% of Turf Decrease)		
Percentiles of explanatory variables	Price, \$	Turf, sqft	20% price increase	30% turf decrease	Sum of water decrease, %	30% price increase	20% turf decrease	Sum of water decrease, %
Min	1.1	0	-2.98 %	0 %	-2.98 %	-4.48 %	0 %	-4.48 %
10%	1.18	0	-3.20 %	0 %	-3.20 %	-4.80 %	0 %	-4.80 %
25%	1.91	0	-5.18 %	0 %	-5.18 %	-7.77 %	0 %	-7.77 %
50% (Median)	2.1	20	-5.70 %	-0.17 %	-5.86 %	-8.55 %	-0.11 %	-8.66 %
Mean	2.33	202	-6.32 %	-1.70 %	-8.02 %	-9.48 %	-1.13 %	-10.61 %
75%	2.99	249	-8.11 %	-2.09 %	-10.20 %	-12.17 %	-1.39 %	-13.56 %
90%	3.1	643	-8.41 %	-5.40 %	-13.81 %	-12.61 %	-3.60 %	-16.21 %
Max	4.52	8115	-12.26 %	-68.14 %	-80.40 %	-18.39 %	-45.43 %	-63.82 %
Coefficients	-0.1211	0.0003	Average, %		-16.21 %	Average, %		-16.24 %

We see that 1<sup>st</sup> and 2<sup>nd</sup> scenarios give on average the same percentage of decreased water use, but effects differ through the price and turf size groups (2<sup>nd</sup> scenario is more evenly distributed among groups)



# CONCLUSIONS

1. Modeling approach relies on use of a fully disaggregated data set
2. The model estimates are robust and support current empirical as well as theoretical knowledge of water demand
3. Results used to create tools for water demand forecasting and targeting the most sensitive groups

**THANK YOU!**

**QUESTIONS?**