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Seasonality and Weather Effects on Electricity Loads: Modeling and Forecasting

Christian Crowley and Frederick L. Joutz

Department of Economics and Research Program on Forecasting

http://www.gwu.edu/~forcpgm



Seasonality and Weather Effects on Electricity Loads: Modeling and Forecasting Preliminary Results – Comments Welcome

Christian Crowley and Frederick L. Joutz Department of Economics and Research Program on Forecasting The George Washington University Washington, D.C. 20052 (202) 994-4899 e-mail: crow@gwu.edu, bmark@gwu.edu

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Objectives

- Examine the temperature effects on electricity loads within the PJM ISO
- Develop parsimonious specifications for hourly load models
- Test one and two day ahead forecasting capability of models
- Simulate the response to a 2°F increase in mean temperature

Objectives

- The research project focuses on short-term and long-term responses to changing temperature, its impact on generation and consumption technologies, and public health.
- Our focus today is only on short-run hourly consumption during the summer months.
- Our work is preliminary we have a lot to learn and expect to benefit from the MetrixND User Meeting hosts and participants.

Progress and Learning

- Collected hourly load data for the control regions
- Collected hourly weather station data from NOAA
- Developed approach for correcting missing observations and outliers in weather data
- Developed a general-to-specific approach for modeling hourly demand

Progress and Learning

- Focus today is only on BG&E, Baltimore's load region within PJM
- Data problems specifically with weather

Progress and Learning

- The Hourly Demand Model depends on:
 - Autoregressive components
 - Daily and Monthly Effects
 - Temperature and Temperature Squared
- We have estimated "cooling effect" elasticities
- We have performed forecast evaluations
- We have run a higher temperature simulation

BG&E Load Region

- In operation for over 100 years
- 1.1 million electricity customers
- 580,000 natural gas customers
- 2,300 square miles
- 1,200 miles of transmission lines
- 21,000 miles of overhead & ground distribution lines
- Traditional vertically integrated utility until July 1st, 2000
- A subsidiary of the holding company, Constellation Energy Group.

BG&E Load Region





"Seasonality and Weather Effects on Electricity Loads" Crowley and Joutz, GWU Sept. 02

Weather Data

- Hourly weather station data from Baltimore-Washington International Airport, WBAN#724060
- 29,184 total observations
- 868 missing observations (3% of total)
- 8,784 observations for June, July, August, and September in 1998, 1999 and 2000
- There are 155 missing observations (1.8%)

<u>Temperature Sensitivity</u> <u>Variables</u>

We constructed Cooling Degree Day variables for modeling purposes:

 $CDD = \begin{cases} F \\ 0 \text{ for Temperatures below } 72^{\circ}F \end{cases}$



Refer to Table 2, 3, and 4 in Handouts



Refer to Table 2, 3, and 4 in Handouts

The General Load Model

$$\log(Load_{ht}) = \beta_0 + \sum_{i=1}^7 \beta_i Day_{iht} + \sum_{i=1}^3 \delta_i Month_{iht}$$

$$+ \sum_{i=1}^4 \alpha_i Month_{iht} \cdot CDD_{iht} + \sum_{i=1}^4 \gamma_i Month_{iht} \cdot CDD_{iht}^2 + \phi(L) \cdot \log(Load_{ht}) + e_{ht}$$

The autoregressive component includes the previous three hours and the same hour from the previous day:

$$\phi(L) = \phi_1 L + \phi_2 L^2 + \phi_3 L^3 + \phi_{24} L^{24}$$

Interpreting the Coefficients

• The electricity load in hour h on Fridays in July is captured by the expression

 $\beta_0 + \beta_5 + \delta_1 + \alpha_2 * CDD + \gamma_2 * CDD^2$

- The Friday effect comes from the β_5 parameter and the δ_1 parameter captures the difference between July and June.
- The effect for a Tuesday in June is given by $\beta_0 + \beta_1 + \alpha_1 * CDD + \gamma_1 * CDD^2$
- The parameter β_1 represents the difference from Monday in June.

General to Specific Modeling Approach

- Goal is to explain a lot with a little
- The general model has 23 parameters to estimate
- We test 40 hypotheses on the deterministic variables on the temperature sensitivities, and both jointly total 960 hypotheses
- Refer to Handout Table 8 for Description
- Refer to Handout Table 9 for Selection

General to Specific Modeling Approach

- Model Evaluation is based on:
 - Likelihood Ratio Statistics
 - Akaike Information Criteria
 - Schwartz Criteria
- Removal of deterministic effects for days and months reduces explanatory power.
- Level and Quadratic terms for CDD are important, but different on an individual monthly basis
- Symmetry found across months as well as for June/September and July/August

MStat Comparison – Hour 10

Model	Coefs	Err DF	Std Err	AIC	BIC	Log-L
Model0_10	23.00	1191.00	0.01	-8.62	-8.52	3536.52
Model40_10	19.00	1195.00	0.01	-8.63	-8.54	3533.29
Model41_10	17.00	1197.00	0.01	-8.62	-8.55	3529.84

•Version 3.0 of MetrixND allows the modeler to conduct the general-to-specific hypotheis tests that we did "by hand"

- •The first line shows results for our General Model
- •Other lines compare results for our Restricted Models 40 and 41

The Final Load Model

$$\log(Load_{ht}) = \beta_0 + \sum_{i=1}^6 \beta_i Day_{iht} + \beta_7 Holiday_t + \sum_{j=1}^3 \delta_j Month_{jht} + \alpha CDD_{ht} + \gamma CDD_{ht}^2 + \phi(L) \cdot \log(Load_{ht}) + e_{ht}$$

•The intercept captures Mondays in June

•Imposing symmetry across monthly CDD effects does not reduce the model's explanatory power

•A close alternative suggested that there is a distinction between the June/September and July/August combinations

•Exmaple: Hour 14 Model in Handout Table 11

Temperature or CDD Elasticities

• The hourly cooling degree day elasticity in month/season j is defined as

$$E_{Chj} = (\alpha_{hj} + \gamma_{hj} * 2 * MCDD_{hj}) * MCDD_{hj}.$$

- Subscripts C, h, and j refer to cooling, hour, and month/season respectively
- The M in front of MCDD terms indicates that degree day measures are evaluated at their mean values
- Refer to Table 15 in Handouts



Forecast Comparison

- Compare Three Approaches:
 - Individual Hourly OLS Model Forecasts
 - Group OLS Forecasts
 - (Group) Neural Net Forecasts
- Estimate models from June-September 1998-1999 and June 2000
- Forecast July 2000 (31 one day ahead forecasts)

Comparison of MAD



Comparison of MAPE



Comparison of RMSE



Simulation

- Compare the effect of a 2°F increase on loads
- Focus on July 2000
- Predicted values calculated using actual temperature data
- Simulated values using actual temperature plus 2 degrees.







Conclusion

- Base-line study for estimating hourly impacts of cooling needs for one load region in PJM
- Forecast comparison we could not distinguish between the different models
- CDD Elasticities were most pronounced in the late morning and late afternoon

Conclusion

- Simulation with temperature increased by 2°F in July 2000
- Biggest average hourly effect is 1% at Hour 10
- Between Hour 9 and Hour 15 the average effect 0.75% or more
- The Hour 10 effects range to 3.25%

Further Research

- Model Specification
 - Are Quadratic Temperature Effects Enough?
 - Cubic Temperature Effects?
 - Convert from natural logarithms to levels?
- Clean Weather Data from Weatherbank
- Other Weather Variables
 - Cloud Cover
 - Precipitation
 - Wind Speed
 - Other Seasonal differences
- Extend modeling efforts to other PJM load regions

Further Research

- MetrixND issues and ideas:
 - Version 3.0 is big improvement over 2.6
 - We relied on econometric package Eviews for many specification tests, because of programming requirements
 - Specification tests compare with MSTAT
 - Elasticity estimates
 - Conversion from transformations to original series
 - Understand how to construct recursive group forecasts in MetrixND
 - Visual Basic for programming and testing large groups of models