

Weather Variability on Store Performance: Improving Resilience for Retail Businesses

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The Opportunity

- Largest private sector employer contributing \$2.5 trillion to annual U.S. GDP (1/5 of the economy)
- Sales fluctuate due to disruptors like financial crises, wars, natural disasters, and weather
- *The Effects of Weather on Retail Sales* by Martha Starr-McCluer (2000) found weather immediately effects consumer spending and can explain changes in spending during future periods

My analysis was able to predict within \$300 of actual store volume

Personal Store Performance

Store Performance By Segment

Plan	LY Act	Plan	Plan	Plan	Plan	LY Act	LY Units	Plan	LY Act	LY Act	LY Act
1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000
1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000	1000000

Application: Meeting User Needs

- Investigate weather impacts on customer behavior and worker productivity
- Utilize collected business metric data (conversion, ADS, traffic) and climate data from store location (Lancaster, PA)
- Perform a set of multiple linear regression analyses to serve as a predictive process in forecasting daily volume
- Provide a risk assessment for store management to incorporate into their daily focus

Analysis

Regression Model: $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_k X_k$

- | | |
|--|--|
| BUSINESS METRICS (PREDICTED VARIABLES) | WEATHER PARAMETERS (PREDICTOR VARIABLES) |
| • Customer-Controlled | • CDD |
| • Traffic | • HDD |
| • Seller-Controlled | • Average Dew Point Temperature |
| • Conversion | • Average Humidity |
| • ADS | • Average Wind Speed |
| | • Average Visibility |
| | • Total Liquid Precipitation Amount |
| | • Total Snowfall Amount |

Implementation: April 10, 2015

Forecasted Weather Variables:

CDD	HDD	Dew Pt.	Humidity	Visibility	Wind	Precip.	Snowfall
0	8	52	83	5	11	0.56	0.00

Predicted Outcomes:

Conversion	ADS	Traffic
16.9%	71.94	478

Calculated Volume = \$5,811.46

Outcome

- Company Planned Volume: **\$8,839.00**
- Predicted Volume: **\$5,811.46** (-34% to Plan)
- Business Strategy
 - Increase conversion to 18%
 - Maximize sales & Maintain ADS
 - Reduce spending by cutting payroll hours by 30% (12 hours)
- Actual Volume: **\$6,094.29** (-31% to Plan)
- Reduced losses by 10%
- Effectively managed payroll
 - Flexibility

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How can we meet users needs?

Many problems in science and engineering require fitting data to a model or fitting equation. The fit can be performed with certain assumptions about the model.

A linear fit assumes that the fitting equation, $y = f(x)$, can be expressed as a linear operator – that is

$$Y = GA$$

This is also called regression and least squares fitting.

Where G is a matrix that describes the fitting equation for the set of independent data points, $x(i)$, that corresponds to $y(i)$. In other words, G is our forward model.

A is a vector that contains the coefficients that we want to solve for.

Eigenvalue Decomposition and Principal Components

- ▶ The German word *eigen* means "self-" or "unique to", or "belonging to".
- ▶ You can model a set of data using these unique eigenvalues as opposed to using an entire data set
- ▶ We can extract the significant numbers eigen-functions from our data using principal components
- ▶ Then we can effectively remove the noise in the observations
- ▶ We then can find a regression operator to statistically predict the coefficients of a model.
- ▶ Amazingly, we do not need to know any of the physics that relate the measurements to the parameters of interest.

Conclusion

- ▶ Cuts down processing power
- ▶ Saves time
- ▶ Saves money
- ▶ Ignoring noise, making data extremely accurate

THANK YOU!