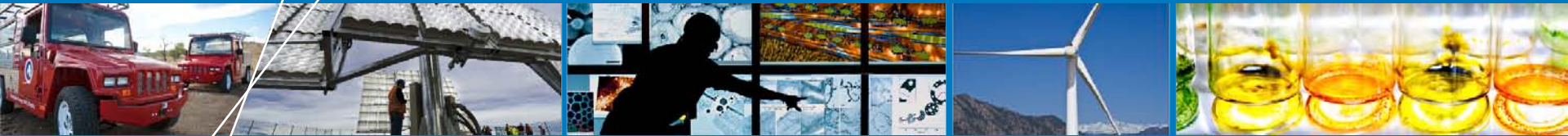


Restaurant Energy Performance Evaluation: How-To Guide and Spreadsheet



Commercial Building Energy Alliance Webinar

Kristin Field

November 16, 2011

Overview

- **What?**

- Resources for comparing restaurant energy performance to that of similar buildings within a portfolio
- How-To Guide
http://apps1.eere.energy.gov/buildings/commercial_initiative/resource_database/detail.cfm?p=367
- Spreadsheet
http://apps1.eere.energy.gov/buildings/commercial_initiative/resource_database/detail.cfm?p=368

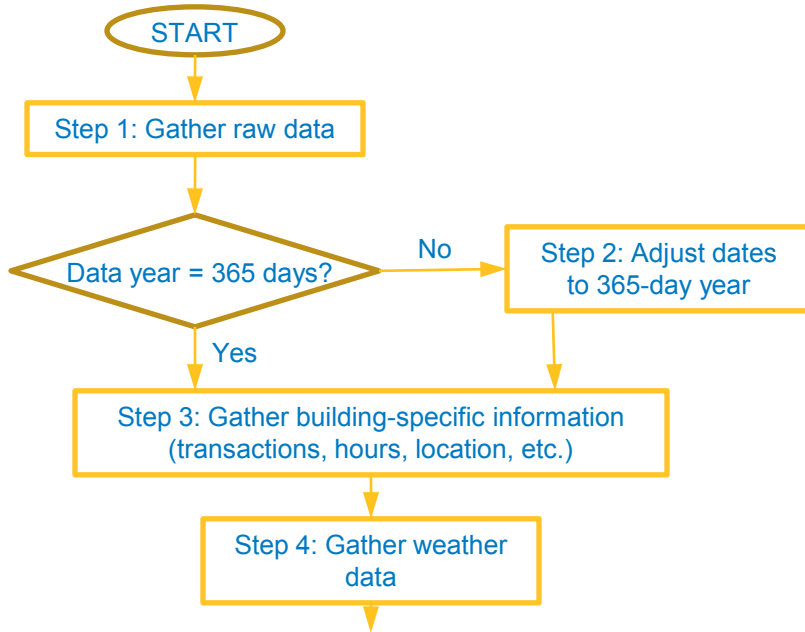
- **Why?**

- Lack of benchmarking guidance for restaurants
 - If you don't know how much energy a restaurant is supposed to consume, how do you know when it's consuming too much?
- Help with retrofit prioritization

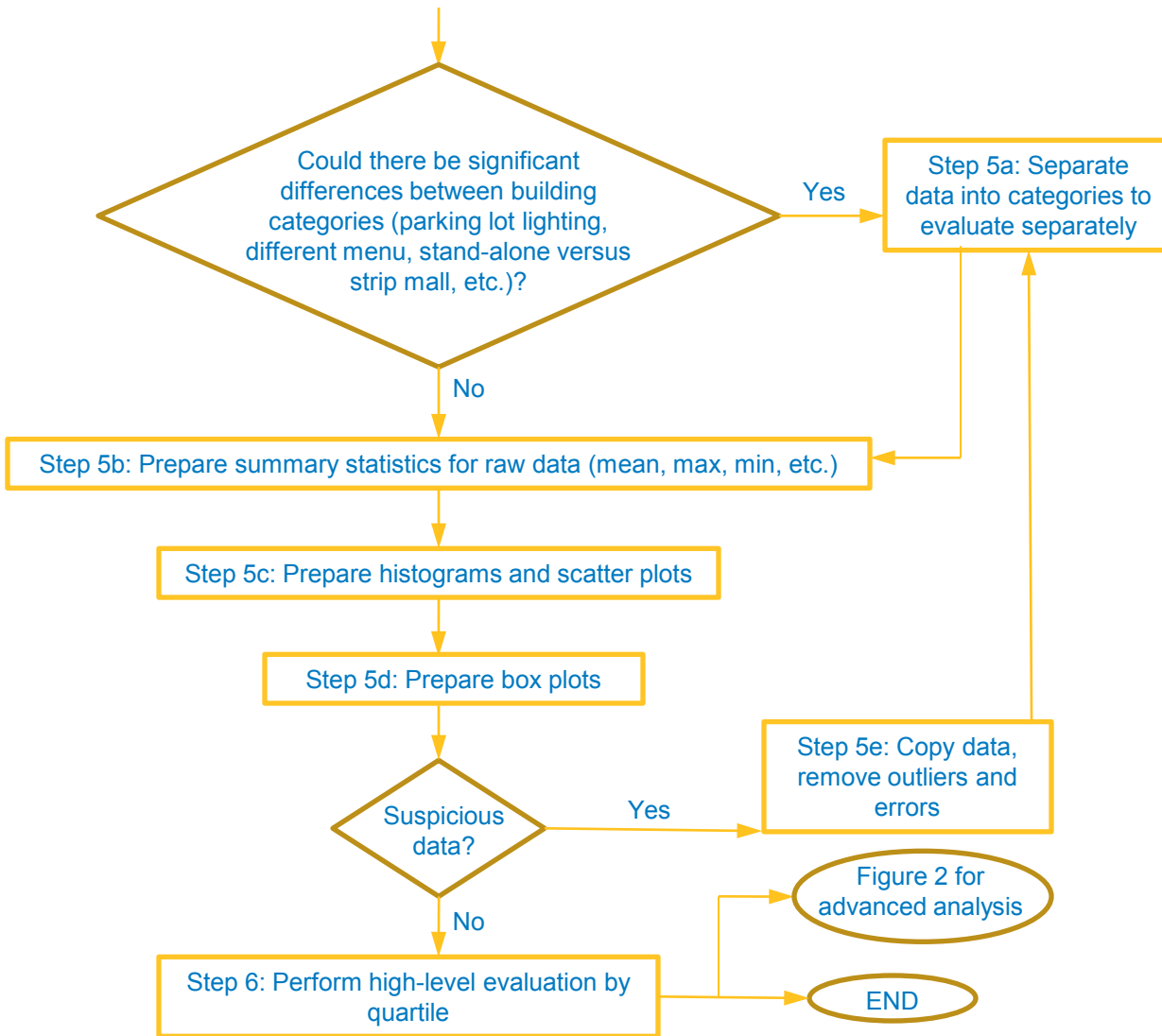
- **How?**

- Document outlines process in steps
 - 6 steps for high-level evaluation
 - 10 steps for advanced evaluation
- Spreadsheet gets calculations started and shows an example from real data

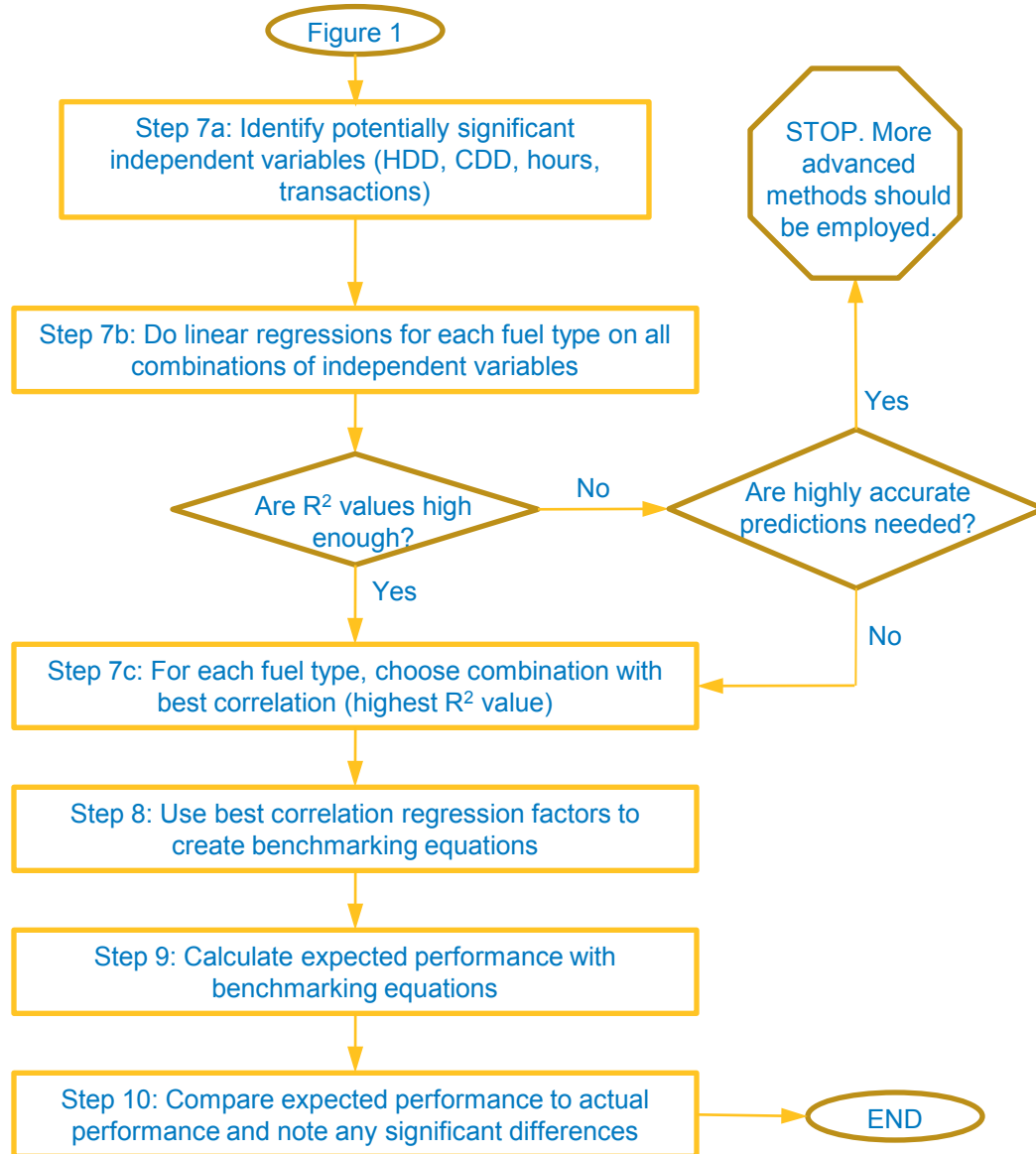
Benchmarking Process Flowchart #1

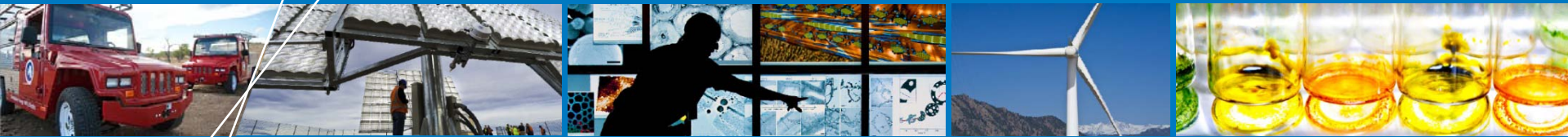


Benchmarking Process Flowchart #1 (continued)



Benchmarking Process Flowchart #2





High-Level Evaluation

Step 1: Gather Raw Data

- Gather raw data:
 - Utility consumption
 - Electricity
 - Natural gas
 - Propane
 - Water/sewer
 - Utility costs

| Weekly Hours | Normalized Weekly Transactions | Sq Ft | Weekly Hours | CDD65 | HDD50 | Annual Electric Usage (kWh) | Annual NG/Propane Usage (therms) |
|--------------|--------------------------------|-------|--------------|-------|-------|-----------------------------------|----------------------------------|
| | | | | | | See General Instructions Step H-v | |
| 121 | 0.3709 | 2781 | 121 | 1294 | 27 | 289,200 | 9,773 |
| 168 | 0.6443 | 4200 | 168 | 1299 | 15 | 362,513 | 12,845 |
| 168 | 0.4816 | 4008 | 168 | 1299 | 15 | 295,711 | 13,236 |
| 126 | 0.5425 | 2759 | 126 | 404 | 56 | 289,488 | 11,154 |
| 126 | 0.5253 | 2968 | 126 | 1506 | 10 | 275,280 | 10,108 |
| 133 | 0.4378 | 4003 | 133 | 1903 | 180 | 280,663 | 11,210 |
| 156 | 0.6403 | 3333 | 156 | 1186 | 14 | 315,615 | 12,650 |
| 168 | 0.4412 | 2545 | 168 | 1299 | 15 | 290,800 | 13,130 |
| 112 | 0.2141 | 1967 | 112 | 1294 | 27 | 193,875 | 8,513 |
| 126 | 0.3876 | 3198 | 126 | 1735 | 613 | 243,216 | 11,545 |
| 126 | 0.3710 | 2192 | 126 | 543 | 32 | 197,157 | 9,581 |
| 133 | 0.3219 | 4083 | 133 | 1528 | 424 | 222,720 | 12,031 |
| 122 | 0.2900 | 2112 | 122 | 1294 | 27 | 178,400 | 8,955 |
| 126 | 0.3620 | 2112 | 126 | 1294 | 27 | 231,920 | 10,289 |

Step 2: Adjust Dates to 365-Day Year

- **365-day year length**
 - Standardizes analysis period
 - Corresponds to year lengths used in external studies, if applicable
- **Most utility records not kept in 365-day periods**
- **Suggested adjustment: add/subtract average daily values from first and last months**

Step 3: Gather Building-Specific Information

- **Building information**
 - Data you think are relevant to restaurant building performance
 - Spreadsheet example uses:
 - Transactions (normalized for anonymity)
 - Hours of operation
 - Floor area
- **Used mostly for advanced analysis (steps 6–10), but useful to gather up front**
- **Potentially used in Step 5a**

Step 4: Gather Weather Data

- **Annual weather data for each location**
- **Suggested metrics, used in spreadsheet example:**
 - HDD50
 - CDD65
- **Potentially time consuming for large portfolio**
 - Normal degree data are a less time-consuming, but less accurate, alternative

Step 5: Choose Subdatasets

- **Step 5 separated into 5 substeps**
 - Step 5a: Separate data into categories
 - Step 5b: Prepare summary statistics for raw data
 - Step 5c: Prepare histograms and scatter plots
 - Step 5d: Prepare box plots
 - Step 5e: Remove outliers
- **May need to iterate all substeps until plots and summary statistics show datasets with:**
 - More linearity
 - Less scatter
 - Few outliers

Step 5a: Separate Into Categories

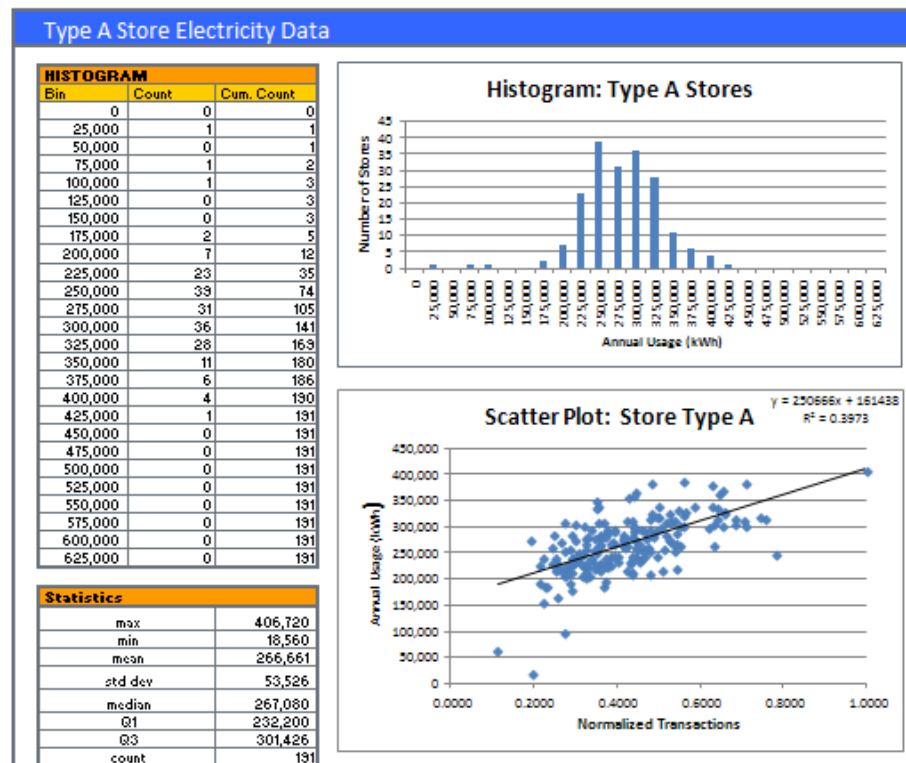
- **Separate into subdatasets by**
 - Store type
 - Stand-alone
 - Strip mall
 - Food court
 - Etc.
 - Inclusion of parking lot lighting in use
 - Menu type
 - Other relevant factors
- **Each subdataset should have at least 50 stores for statistical significance**

Step 5b: Prepare Summary Statistics for Raw Data

- **Examine raw data using summary statistics**
 - By category (from Step 5a) and all together
 - Summary statistics include:
 - Maximum
 - Minimum
 - Mean (average)
 - Median (50th percentile)
 - Standard deviation
- **Do summary statistics support categories chosen in Step 5a?**
 - Are maximum and minimum reasonable numbers?
 - Is standard deviation reasonable?

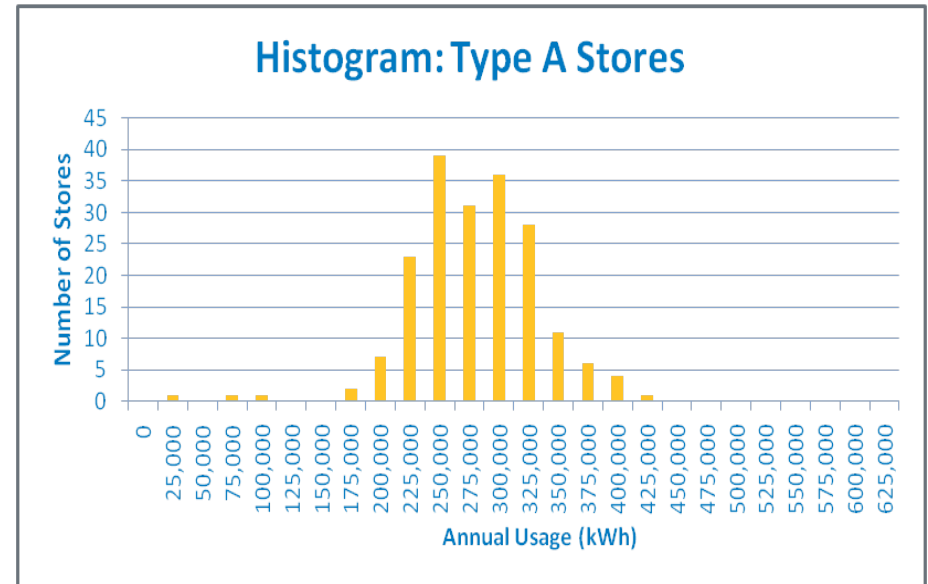
Step 5c: Prepare Histograms

- Show entire distribution of data, instead of just significant points
- Automatically generated by spreadsheet tool



Step 5c: Prepare Histograms (continued)

- **Histograms display:**
 - Y-axis: count of stores (this case) falling within bins shown on x-axis
 - X-axis: bins for variable of interest (electricity consumption in example)
 - Less confusing if bins are all equally sized
- **Histograms can tell you:**
 - Rough numbers to expect for most stores
 - How many outliers
 - Whether values are widely spread or tightly clustered



Step 5c: Prepare Scatter Plots

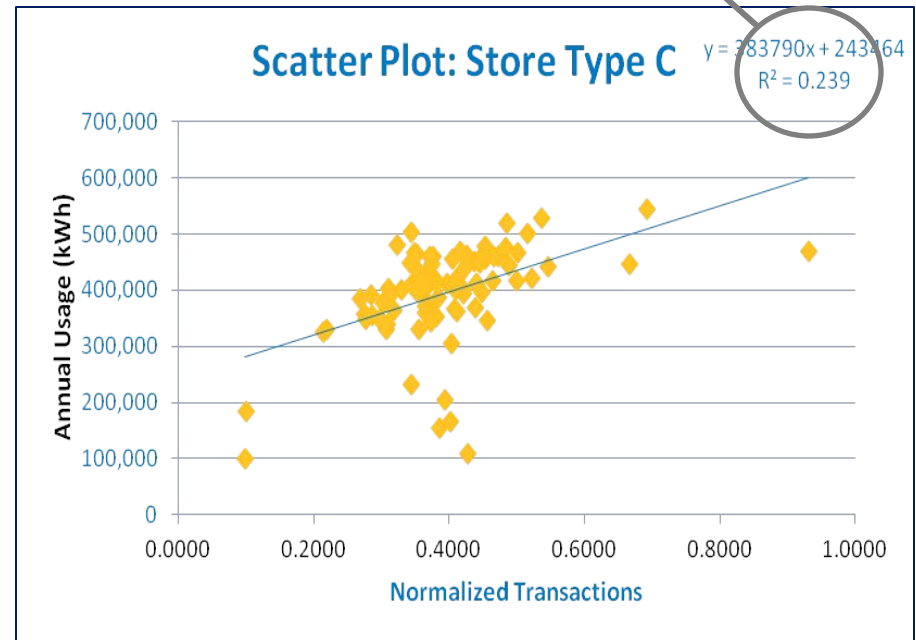
- **Scatter plots display:**

- Y-axis: individual data point values (electricity consumption in example)
- X-axis: individual data point values (normalized transactions in example)
 - Could be helpful to try different variables on x-axis

- **Scatter plots can tell you:**

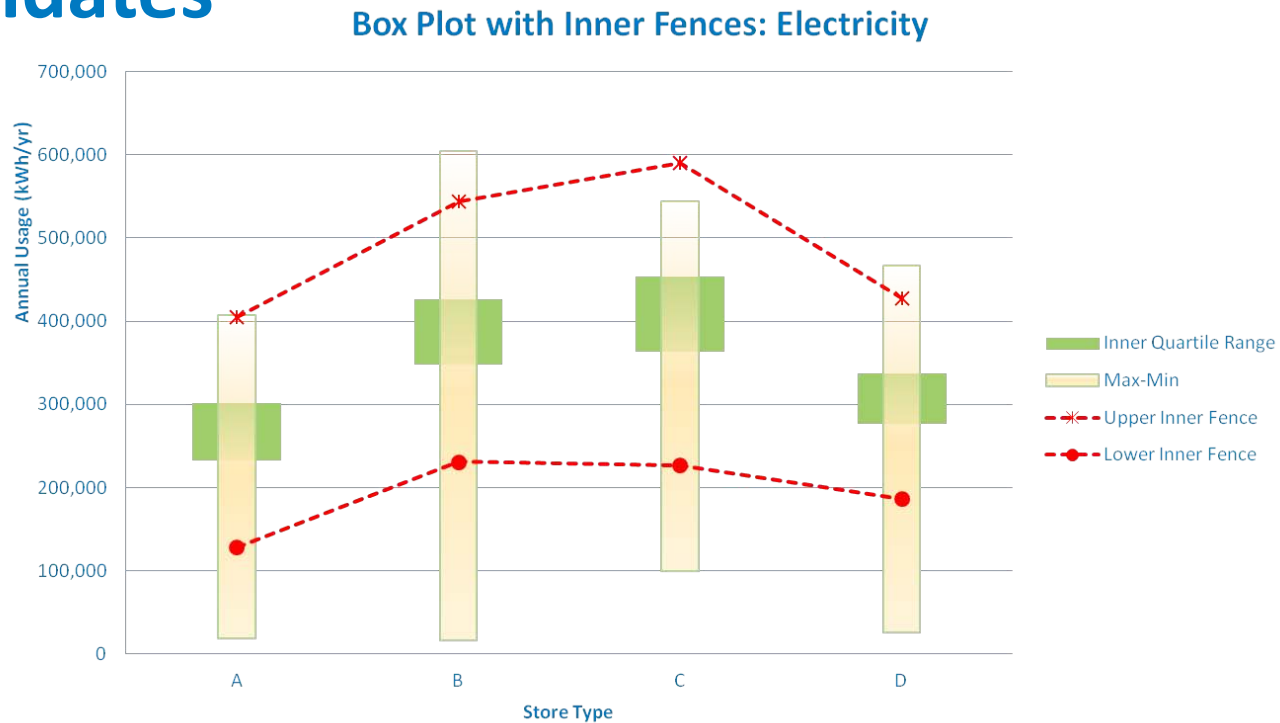
- Rough numbers to expect for most stores
- How many outliers
- Whether values are widely spread or tightly clustered
- Does x-axis value seem to affect y-axis value?
- How strong is the correlation? (R^2 value with Excel trendline)

Example of low R^2



Step 5d: Prepare Box Plots

- Useful display for data in middle of distribution
- Visually identifies preliminary retrofit candidates



Step 5e: Remove Outliers

- Steps 5a–5d help identify outliers
- Spreadsheet also highlights potential outliers in gray on “Complete Inputs” tab
- Mild or extreme
 - Remove mild outliers?
- Outliers caused by:
 - Missing data
 - Additional data (double month)
 - Problem with meter
 - Renovations
 - Other unusual activity
- After removing outliers, repeat steps 5a–5d to see if statistics/plots have improved
 - Reasonable numbers
 - Scatter
 - Skew

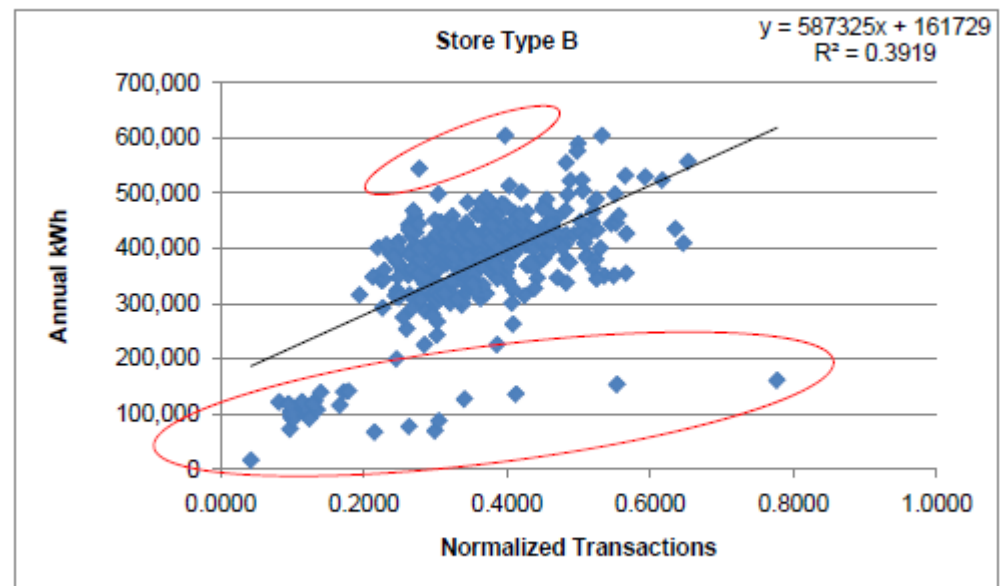
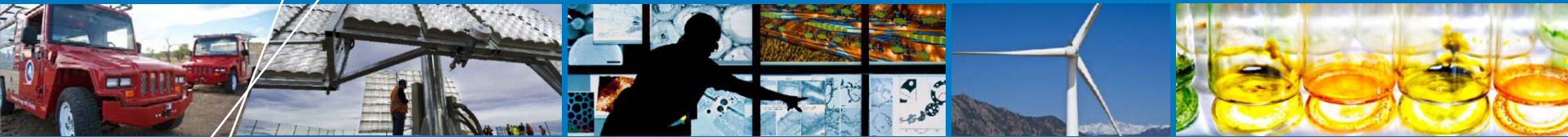


Figure 2-5 Example of annual electricity use scatter plot to show outliers

Step 6: Perform High-Level Evaluation

- **Gather and review information collected in Steps 1–5**
 - Are store categories significant? (Step 5a: stand-alone versus food court, etc.)
 - Range of performance (max–min)
 - Average performance
 - Distribution of performance
 - Scatter?
 - Skew? If so, high or low?
 - Many or few outliers?
- **Identify preliminary retrofit candidates**
 - Spreadsheet highlights in yellow on “Complete Inputs” tab
- **Decision:**
 - Stop after Step 6 and investigate preliminary retrofit candidates, or
 - Continue to Step 7 for advanced evaluation?
 - Develop benchmarking equations
 - Predict energy use given different operation
 - Strong correlations not guaranteed (may invest a lot of time for little confidence in results)



Advanced Evaluation

Step 7: Perform Linear Regressions

- **Step 7 separated into 3 substeps**
 - Step 7a: Identify significant variables
 - Step 7b: Perform linear regressions
 - Step 7c: Evaluate regression quality
- **Spreadsheet completes task automatically in “Regression Analysis” tab**

Step 7a: Identify Significant Variables

- **Which factors (independent variables) may be significant in predicting performance?**
 - Example includes:
 - Transactions (normalized for anonymity)
 - Weekly hours of operation
 - Floor area
 - Weather (HDD50, CDD65) for “normal” year
- **Performance by store type is typically electricity and natural gas consumption**

Step 7b: Perform Linear Regressions

- Perform linear regression for each store type
- Example uses Excel's LINEST function
- Regression equation type example:

$$\text{kWh} = a + b * (\text{HDD50}) + c * (\text{Floor Area})$$

where:

a = constant

b = HDD50 slope

c = floor area slope

Step 7c: Evaluate Regression Quality

- **How well does regression predict performance?**
 - High R^2 value (greater than 70%–80%)
 - Other statistical parameter (P-factor, F-test)
 - Commercial statistical software output
- **For best results, evaluate strength of regressions using a variety of combinations of independent variables**

| Independent Variable | Store Type |
|--|------------|
| Transactions, Weekly Hours, Floor Area | A |
| | B |
| | C |
| | D |
| Transactions, Floor Area | A |
| | B |
| | C |
| | D |
| Transactions | A |
| | B |
| | C |
| | D |
| Floor Area, Weekly Hours | A |
| | B |
| | C |
| | D |
| Weekly Hours | A |
| | B |
| | C |
| | D |
| Floor Area | A |
| | B |
| | C |
| | D |
| Weekly Hours, Transactions | A |
| | B |
| | C |
| | D |

Step 8: Create Benchmarking Equations

- Choose strongest regression correlation
- Using the coefficients corresponding to the strongest correlation, create benchmarking equations for:
 - Each fuel type (electricity, natural gas, etc.)
 - Each store type (stand-alone, strip mall, etc.)

- **Example:**

$$\text{kWh} = a + b * (\text{Transactions}) + c * (\text{Floor Area}) + d * (\text{Weekly Hours}) + e * (\text{CDD}) + f * (\text{HDD})$$

where:

a = constant

b,c,d,e,f = regression coefficients (“slopes”)

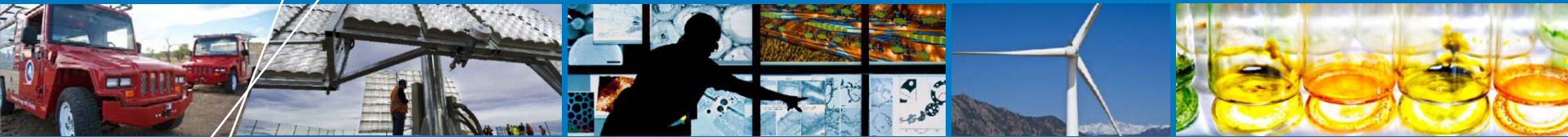
| | | Electricity | | | | | | |
|---|------------|-------------|--------------|------------|--------------|-------|-----------|----------------|
| | | Intercept | Transactions | Floor Area | Weekly Hours | CDD | HDD | R ² |
| Independent Variable | Store Type | Slope | Slope | Slope | Slope | Slope | Slope | |
| Transactions, Weekly Hours, Floor Area, CDD, and HDD | A | -45,829.4 | 1.3 | 18.7 | 1,270.2 | 18.2 | 135,634.9 | 0.61 |
| | B | 187,551.8 | -21.2 | -8.7 | -400.4 | 18.1 | 611,678.8 | 0.42 |
| | C | 517,680.3 | -59.1 | -66.7 | -328.5 | -11.5 | 93,694.2 | 0.34 |
| | D | 906.0 | -0.2 | 24.3 | 756.9 | 22.9 | 33,692.0 | 0.48 |

Step 9: Calculate Expected Performance

- **Use benchmarking equations to calculate expected performance of each store**
 - Performance = y values
 - Given known operational characteristics = x values
 - Transactions
 - Hours of operation
 - Floor area
 - Weather
 - Etc.
 - Calculate y , given x 's

Step 10: Compare Expected Versus Actual Performance

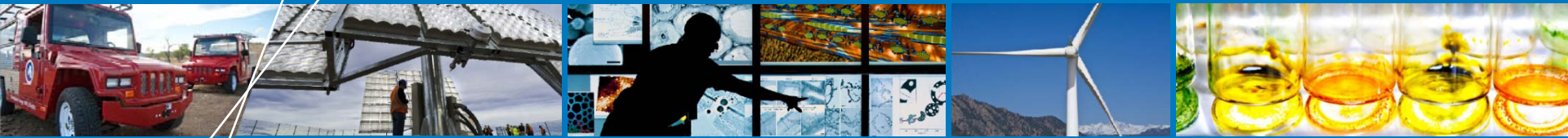
- **Compare expected performance (from Step 9) to actual performance (from collected data)**
 - Comparison in absolute numbers (e.g., kWh or therms)
 - Comparison in % usually more helpful
- **General guidelines for interpretation**
 - Less than 5% variation: in the noise
 - 5%–10% variation: note deviation
 - 10%–25% variation: consider energy audit
 - Greater than 25% variation: probably data error



Next Steps

Next Steps

- **If any values have >25% variation, remove them as outliers and iterate calculations**
- **Look into stores with 10%–25% variation**
 - May warrant energy audit to help determine whether retrofits make sense
 - May indicate mild outliers
 - May have reasonable explanation for higher consumption
- **Any necessary customization of spreadsheet tool to meet the needs of an individual portfolio**
- **Feedback to DOE/NREL**

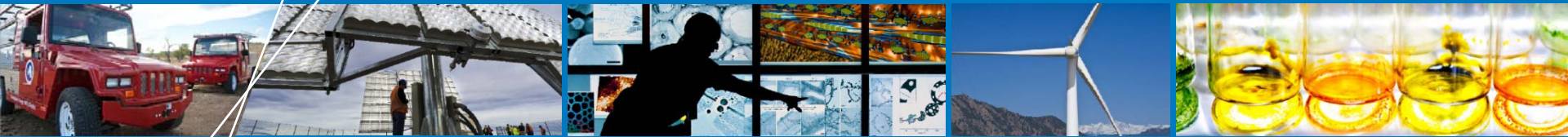


Acknowledgments

Acknowledgments

NREL gratefully acknowledges the contributions of the following individuals:

- Vernon Smith
- Roger Hedrick, Architectural Energy Corporation
- Rachel Romero, NREL
- Members of the REA-Restaurant Project Team



Questions?

**Contact information:
Kristin.Field@nrel.gov**