



Digital Egypt Pioneers Initiative - DEPI

Electricity Group 2 (Eng. Sherihan Ali)

Project Administrators

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Table of Contents

Additional Columns	3
Assumptions and Remarks	3
Date and Time	3
Data Status	4
Demand Loss (MW)	4
Event Type	5
Detailed Event Type	5
Cleaning Steps	6
Date and Time	6
Area Affected	7
NERC Region	7
Event Type	8
Natural Disaster Type	8
Demand Loss (MW)	8
Number of Customers Affected	8
Key Performance Indicators (KPIs)	9
1. ImpactScore	9
2. Total Customers Affected	9
3. Average Event Duration	10
4. Total Demand Loss (MW)	10
5. Number of Events by Type	10

Additional Columns

We added some columns to enhance the dataset, providing more detailed insights into the power outage events. These new columns are:

Duration Time

This column captures the duration of the outage, calculated by subtracting the Date Event Began from the Date of Restoration.

Detailed Event Type

This column provides more specific information about the event categories, which will be explained later in the report.

Data Status

This column indicates the type of data, categorized as Faulty, Anonymous, Full, or Fair. Further explanations of these classifications will be provided later in the report.

Years Group

This column organizes the data into specific year ranges, grouping events by periods such as 2002-2010, 2011-2014, 2015-2022, and 2023. This helps in analyzing trends and patterns over time.

Assumptions and Remarks

Date and Time

The time **5:78 PM** in the 2005 sheet was identified as a typographical error. This occurred due to the proximity of the numbers 7 and 4 on the keyboard, leading to an incorrect entry.

Converted the day's events:

Evening: 6:00 PMNoon: 12:00 PMMidnight: 12:00 AM

In cases where the date and time of restoration are unknown and the Demand Loss (MW) is recorded as zero, we assumed that the electric current was restored simultaneously. Therefore, the date and time of restoration will be considered the same as the recorded time and date of return.

Data Status

The input is considered **Faulty** if the power outage duration exceeds one hour and the Demand Loss (MW) is recorded as zero.

The data is considered **Anonymous** if the Number of Customers Affected and Demand Loss (MW), the event start date, and the restoration date are NULLs.

The data is considered **Full** Data if the Number of Customers Affected, Demand Loss (MW), Duration Time, and NERC Region are not NULLs.

Otherwise, It is considered **Fair** Data.

Demand Loss (MW)

We assumed that **none** is equivalent to **zero** during data processing.

For (PG&E) values: Assumptions used in calculating the impact of power outages:

- 1. Average household consumption: Assumes each household consumes 1.2 kWh per hour according to official data of NERC.
- 2. Outage duration: Calculated as the difference (in hours) between the Time Event Began and the Time of Restoration.
- 3. Number of individuals per household: Assumed to be 3 individuals per household.
- 4. Equal energy distribution: The lost energy is assumed to be distributed equally among the affected households.

We used AI to help us get the Demand Loss value for these two records:

- 1. 133 on 5/21/04 between 3:00 a.m. and 4:00 a.m., 392 on 5/21/04 between 4:00 p.m. and 5:00 p.m.
- 2. 177 on 5/21/04 between 3:00 p.m. and 5:00 p.m.

Using the following assumptions that it gave to us:

- Energy loss during the specified time total_loss = loss_3_to_4_am + loss_4_to_5_pm
- Assume the remaining hours have a similar loss (average of the known periods)

remaining_hours = duration_hours - 2

- Subtract the two known periods average_loss = total_loss / 2
- > The average loss of the two known periods
- > Total energy lost total_energy_lost = total_loss + (remaining_hours * average_loss)

So the results are:

```
1st record's total_energy_lost = 18,375 MW
2nd record's total_energy_lost = 6,195 MW
```

If there was a peak value and an accumulated value, we took the peak value.

Event Type

In the 2023 sheet, the most recently updated file for all sheets, we noted that "Public Appeal" refers to "Shedding Load". Therefore, we considered it equivalent to the previous years as well.

Detailed Event Type

This column categorizes events into more specific types, which will be explained as follows:

Natural Disaster

- Severe Storms
- Winds
- Snow/Ice Events
- Heat Wave
- Earthquakes
- Floods

Fire

o No specific categories

Vandalism

- Physical Attack
- Cyber Attack
- Suspicious Activity

• Operational Malfunction

- Shedding Load
- o Inadequate Resources
- System Issues

- Network Issues
- Equipment Failure

Others

Any event types not covered by the above categories.

Cleaning Steps

We divided the sheet into groups: **2002-2010**, **2011-2014**, **2015-2022**, and **2023**, ensuring that each group maintains the same structure. After cleaning the date and time columns, we will append these groups into a single query. Subsequent cleaning steps will then be applied.

We added this record manually to the 2003 sheet as it was grouped to another record and deleted while cleaning

27| 2/5/2003 SERC 12,897 (Alabama) 8:00 p.m. Alabama Severe Thunderstorms 130 12,897 (Alabama) 5/03/03, 8:00 a.m.

We didn't delete any records except the Blanks Rows and the Duplicated ones which are 15 records so the final total records is **3936** after it was **3951**.

Basic Cleaning Steps for Each Group

- 1. Remove Blank Rows
- 2. Remove Top Rows
- 3. Promote Headers

We utilized functions and custom columns to optimize performance and reduce file load during cleaning. To streamline the workflow, we focused on applying **dynamic** changes rather than manual adjustments.

Now, let's examine what has occurred in each dataset column. We will focus on identifying trends, inconsistencies, and any necessary adjustments made during the cleaning process.

Date and Time

Replaced wrong data like

- 7/01//05 in the 2005 sheet with 1/7/2005
- Text values like "Unknown", "Ongoing", "NA", and "(Trans. Only)" with NULL
- Wrong input data like the year 2024 in the Date of Restoration in the beginning data 2006-2010

18/3/2001 and 29/8/2077 in the 2011 sheet with 18/3/2011

Area Affected

When we placed the **Area** column on a map visualization, we observed that some entries, such as "Eastern Montana" and "Vallee, California," were incorrectly located outside the USA, specifically in Asia and Australia. As a result, we corrected these entries to ensure accurate geographic representation.

NERC Region

The **North American Electric Reliability Corporation (NERC)** divides the U.S. and parts of Canada into six major regional entities, each responsible for overseeing the reliability of the power grid in their area.

Below are the main NERC regions along with their abbreviations and the territories they cover:

- 1. **Midwest Reliability Organization (MRO)**: Covers parts of the U.S. Midwest and Canada, including Minnesota, Wisconsin, Iowa, and Manitoba.
- 2. **Northeast Power Coordinating Council (NPCC)**: Includes New York, New England, Ontario, Quebec, and the Maritime provinces of Canada.
- 3. **ReliabilityFirst Corporation (RFC)**: Covers the Great Lakes region, including parts of Ohio, Pennsylvania, Maryland, New Jersey, and Virginia.
- 4. **Southeastern Electric Reliability Council (SERC)**: Covers the southeastern U.S., including the Carolinas, Georgia, and parts of Mississippi, Alabama, and Tennessee.
- 5. **Texas Reliability Entity (TRE)**: Focuses on the state of Texas, which operates mostly independently from the national grid.
- 6. **Western Electricity Coordinating Council (WECC)**: Covers the western U.S., including California, Arizona, Nevada, and portions of Canada and Mexico.

These two regions aren't set under the control of NERC but they were in the data:

- 7. **Hawaii** (HI) has its own utility and grid operators, like Hawaiian Electric Company (HECO), which are not part of the continental NERC regions
- 8. **Puerto Rico (PREPA)** has its own utility and grid operators and doesn't fall within the main NERC regions.

Event Type

Multiple event types could be grouped under broader categories. After classification, the final categories are:

- Natural Disaster
- Fire
- Vandalism
- Shedding Load
- Operational Malfunction

Natural Disaster Type

We observed that "Natural Disaster" was the most common category. To gain more insights, we added this new column to classify specific types of natural disasters, such as winds, floods, and earthquakes.

Demand Loss (MW)

We replaced wrong data like

• Text values like "Unknown", "All", and "NA" with NULL

There were numbers separated by **"to"** and **"-"** so we took the average, like "65 to 100" and "8000-10000"

Number of Customers Affected

We replaced wrong data like

 Text values like "Unknown", "utilities", "industrial", and "Interruptible" with NULL

Key Performance Indicators (KPIs)

the key performance indicators (KPIs) are used to analyze power outage events, assess their impact on customers, and identify critical infrastructure improvement and service response optimization areas. The analysis focuses on KPIs such as ImpactScore, Outage Risk Index (ORI), Event Severity Index (ESI), and other metrics, providing actionable insights to support better decision-making and resource allocation.

1. ImpactScore

- **<u>Definition:</u>** Measures the overall impact of power outages by calculating the product of demand loss and the number of customers affected.
- **Purpose:** This KPI will help quantify the overall impact of each outage by combining the lost energy (Demand Loss) and the number of customers affected. By evaluating this metric across events, you can assess which outages had the most significant consequences and prioritize preventive actions.
- <u>Insights:</u> Consistently high ImpactScores in a region indicate an urgent need for infrastructure upgrades or more reliable systems to mitigate future outages in those areas.

• Formula:

ImpactScore = Demand Loss (MW) x Number of Customers Affected

2. Total Customers Affected

- **<u>Definition</u>**: Total number of customers impacted by power outages.
- <u>Purpose:</u> This KPI gives a clear picture of how many customers were impacted across all events. It helps in understanding the broader community impact of outages.
- <u>Insights:</u> A high number of total customers affected across multiple events may indicate that certain regions or customer segments are more vulnerable to outages. This insight can help prioritize grid reinforcements or emergency response plans in areas with a large customer impact.

• Formula:

Total Customers Affected = Σ Number of Customers Affected

3. Average Event Duration

- **<u>Definition:</u>** Measures the average duration of power outages over a defined period.
- <u>Purpose:</u> This KPI measures how long outages typically last, which helps in identifying trends related to recovery times and evaluating the performance of recovery teams.
- Insights: A longer-than-average Event Duration in a particular region may suggest resource allocation issues, such as insufficient response teams or difficulties accessing affected areas. Addressing these issues can minimize downtime in future outages.

• Formula:

Average Event Duration = $\frac{Total\ Duration\ Time}{Total\ Number\ of\ Events}$

4. Total Demand Loss (MW)

- **<u>Definition</u>**: The total amount of demand loss in megawatts during power outages.
- **Purpose:** This KPI sums up the total energy lost across all events, providing a clear view of how much power was disrupted over time.
- <u>Insights:</u> A high Total Demand Loss over several years could suggest that certain regions are more prone to larger disruptions, requiring focused investments in improving grid stability or upgrading systems.

• Formula:

 $Total\ Demand\ Loss\ (MW)\ =\ \Sigma\ Demand\ Loss\ (MW)$

5. Number of Events by Type

- **<u>Definition</u>**: Counts the number of events categorized by their type (e.g., natural disasters, operational failures).
- **Purpose:** This KPI categorizes events by type (e.g., natural disasters, operational failures) to identify which types of incidents occur most frequently and cause the most disruption.
- <u>Insights:</u> By identifying that operational failures happen more often than natural disasters, DOE could invest more in equipment maintenance and training to reduce these outages.

• Formula:

 $Number\ of\ Events\ by\ Type\ =\ Count\ of\ Events\ grouped\ by\ Event\ Type$