

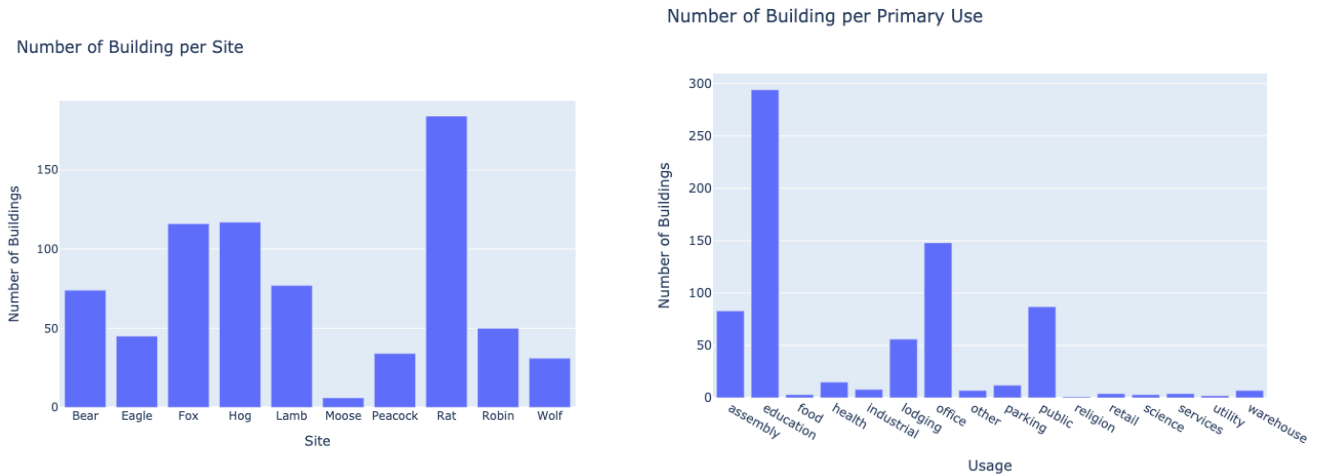
Exploratory Data Analysis (EDA)

After data cleaning, we explored the dataset to understand the distributions and trends in energy consumption. We categorized the meter data by site and primary electricity usage.

Before Data Manipulation

Distribution

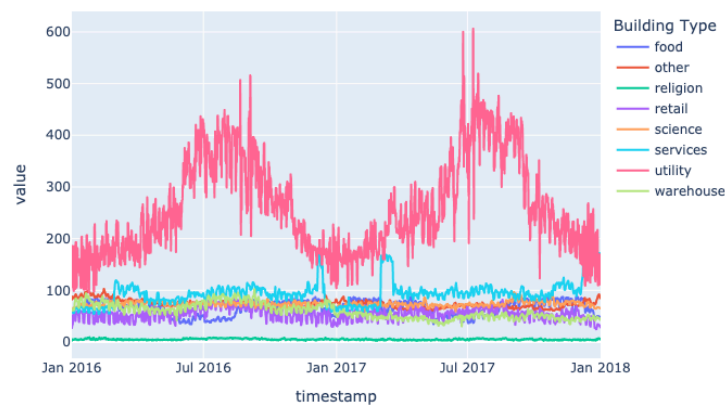
By plotting bar charts,



We observed that there are **Data imbalance** problems occur in both categorization by site and energy primary usage. For this site, “Rat” has the highest number of buildings (184, ~25% data) and Moose has the fewest (6, ~0.77% data). For primary use, “education” is the largest (294, ~40%) data and 8 categories of primary use have less than or equal to 8 buildings (~1.09% data).

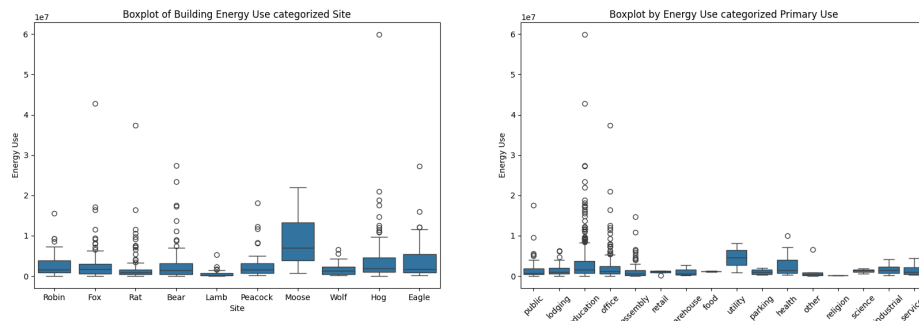
Focused on primary usage categories with ≤ 8 buildings, we investigated whether categories have distinct energy trends to decide whether to combine them into a single group or not.

Mean Energy Use Over Time by Primary Use (24-Hour Rolling Avg)



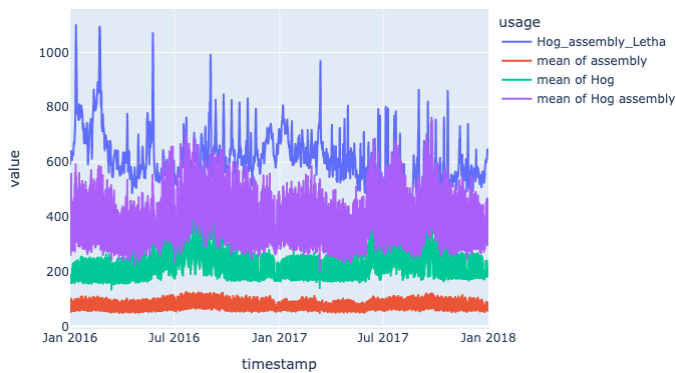
To assess this, we plotted line graphs of the primary use categories with few buildings. We observed that the “utility” category has consistently high energy consumption and decided to keep them as a single category. In contrast, “religion” has consistently low energy use (~ 5 kwh) at all times, making its contribution negligible compared to other categories. Moreover, data in category “services” have sharp increases in energy consumption during January and April, followed by decrease at different times, which does not show an obvious seasonal pattern. Therefore, we decided to (1) keep “utility” as a single category, (2) remove data in the “religion” category and (3) combine the remaining 6 categories, i.e. [], as category “other”.

By plotting box plots,

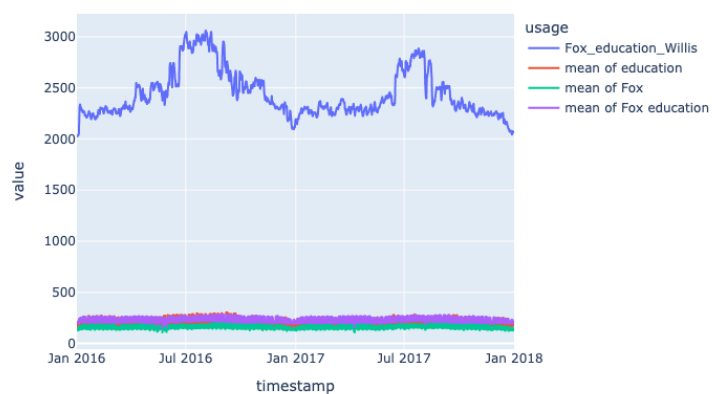


We observed some outliers and decided to look into the time series data of the outliers and compare it with the mean of energy use with the same site and/or primary use.

Mean Energy Use Over Time Hog_assembly_Letha (24-Hour Rolling Avg)



Mean Energy Use Over Time Fox_education_Willis (24-Hour Rolling Avg)



There are 2 kinds of outliers, the first kind of outlier (left side plot) is close to the mean of same primary use /same site and primary use and we classify these data as good to keep. The second kind of outlier is much higher than the means of the same site and/or primary use. As they have high energy use, they may have higher traffic of customers and are critical for us for predicting future energy usage and deriving business insights to reduce energy waste and cost. Also, these outliers with high energy use may have and As a result, we decided to keep these checked data.

Data Manipulation

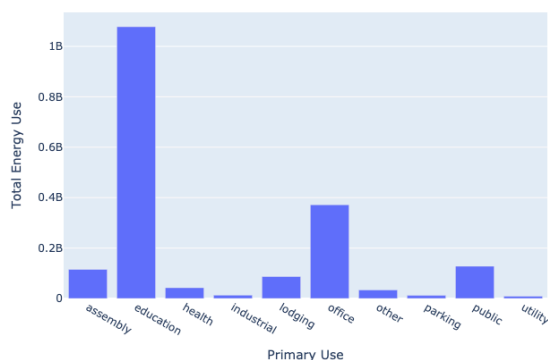
To conclude the above, we decided to remove data in the “religion” category and categories [] as category “other”.

Investigate Data after Data Manipulation

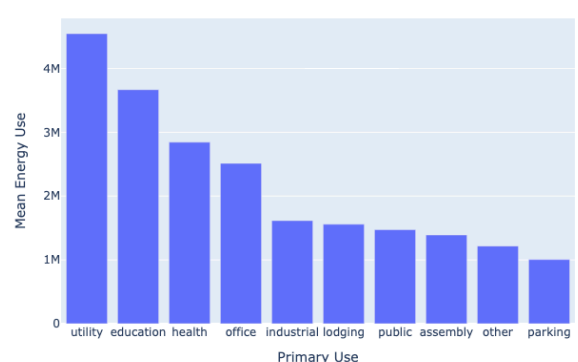
Data Distribution

By plotting the bar charts in terms of primary use,

Total Energy by Primary Use



Mean Energy Use per Building by Primary Use

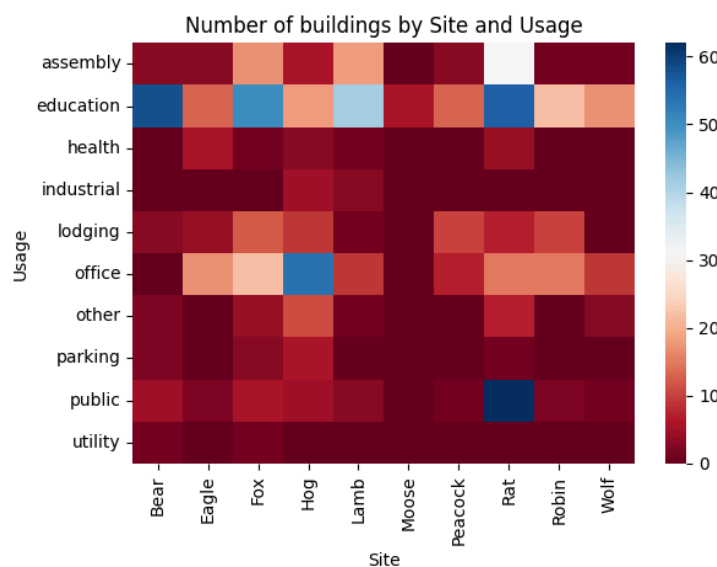


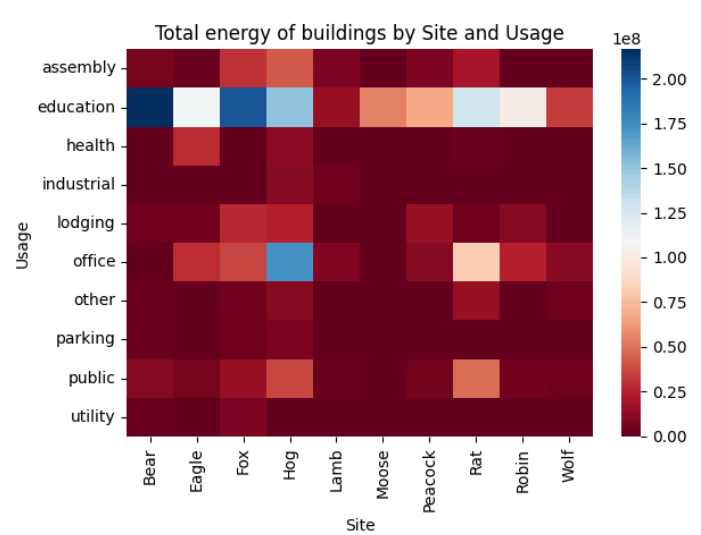
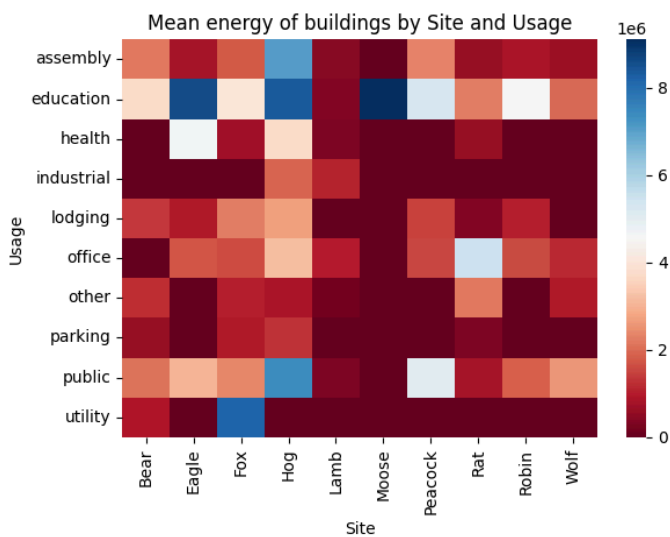
By plotting the bar charts in terms of site,

Site	Mean Energy Use (M)
Bear	3.2
Eagle	4.0
Fox	2.8
Hog	4.0
Lamb	0.5
Moose	9.5
Peacock	3.1
Rat	1.6
Robin	2.8
Wolf	1.6

Site	Total Energy Use (M)
Bear	240
Eagle	180
Fox	320
Hog	460
Lamb	30
Moose	50
Peacock	100
Rat	290
Robin	140
Wolf	50

By plotting heatmap,



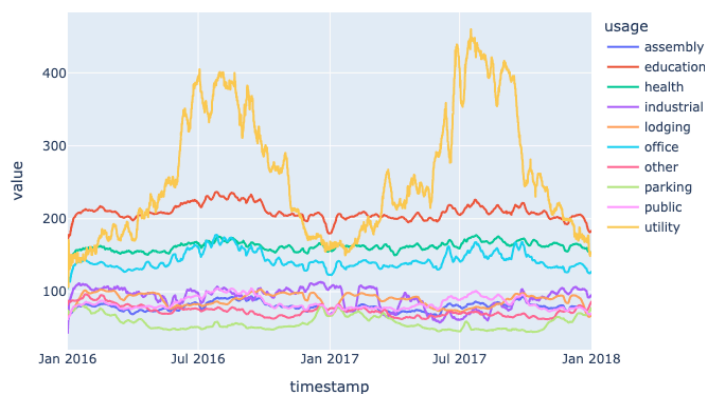


We observed that there are only education buildings on site “Moose”. Also, we observed that “education” buildings in all sites, “health” buildings in “Eagle”, “utility” buildings in “Fox” and “assembly”, “public” buildings in “Hog” havr. Moreover, there are sin

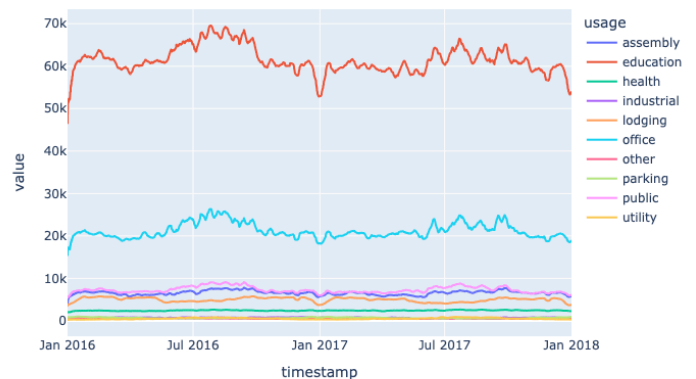
Trend

By plotting line graphs,

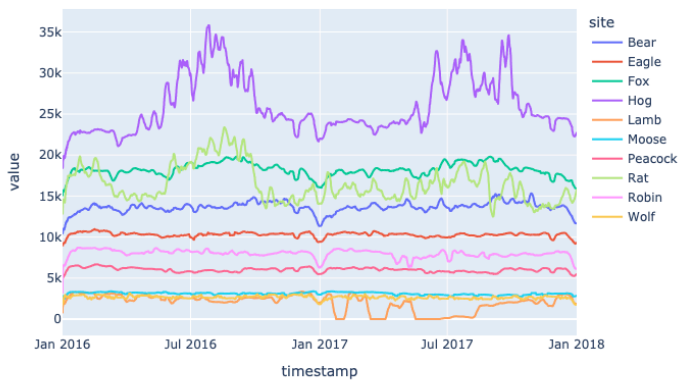
Mean Energy Use Over Time by Primary Use (Weekly Rolling Avg)



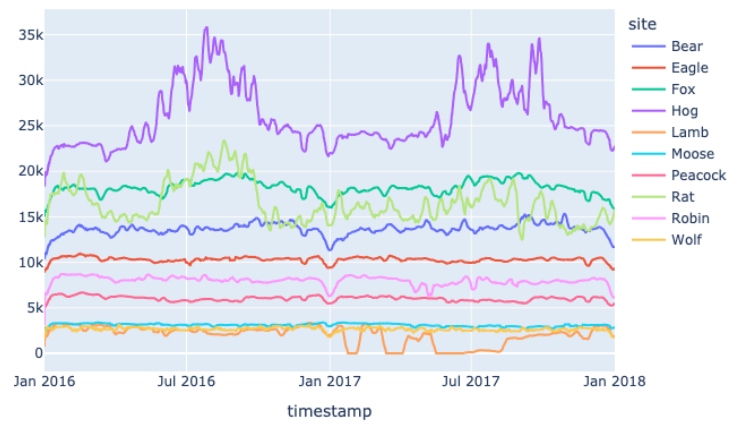
Total Energy Use Over Time by Primary Use (Weekly Rolling Avg)



Total Energy Use Over Time by Site (Weekly Rolling Avg)



Total Energy Use Over Time by Site (Weekly Rolling Avg)



We observed the overall trends, changes of energy usages and We observed that the energy consumption of “utility” follows seasonal or temperature changes.

To conclude, we find seasonal changes in education and utility. Moreover, we observed high mean energy consumption in education, utility and public.