

sprint_1

2023-04-15

Section #01: Benchmarking Data Exploration

Data source for this section

- 2016 LL33 Data Disclosure for CY2015 reporting, Government of New York City. <https://www.nyc.gov/site/buildings/codes/benchmarking.page>

The purpose of this section is to understand the dataset that we are going to use as the input of building energy consumption intensity - 2015 building energy consumption benchmarking data collected under NYC Local Law 84/133 Energy Benchmarking, which requires owners and managers of buildings larger than 50,000 square (25,000 after 2016) to report their building's energy usage to the City of New York on a yearly basis.

The column representing the energy consumption intensity is Site_EUI, and there are many factors impacting this value, like Water, Electricity, building age

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.4.0      v purrr   1.0.1
## v tibble  3.2.1      v dplyr   1.1.1
## v tidyr   1.3.0      v stringr 1.5.0
## v readr   2.1.3      v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
library(ggplot2)
```

```
# Load raw data
```

```
file_path <- file.path(dirname(rstudioapi::getSourceEditorContext()$path), "dataset/NYCBuildingEnergyUs
raw_data <- read_csv(file_path)
```

```
## Rows: 13223 Columns: 57
```

```
## -- Column specification -----
```

```
## Delimiter: ","
```

```
## chr (30): Coreported_BBL_Status, BBLs_Coreported, Reported_BINs, Property_Na...
```

```
## dbl (27): Record_Number, Order, BBL, Street_Number, Zip_Code, Largest_Proper...
```

```
##
```

```
## i Use `spec()` to retrieve the full column specification for this data.
```

```
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
# Pre-processing
```

```
my_data <- raw_data %>%
```

```
  filter(DOF_Benchmarking_Submission_Status == "In Compliance") %>% # filter valid Benchmarking Submiss
```

```
  select(Record_Number,
```

```
    Site_EUI_kBtu_per_sqft,
```

```
    Weather_Normalized_Site_Electricity_Intensity_kWh_per_sqft,
```

```
    Weather_Normalized_Site_Natural_Gas_Intensity_therms_per_sqft,
```

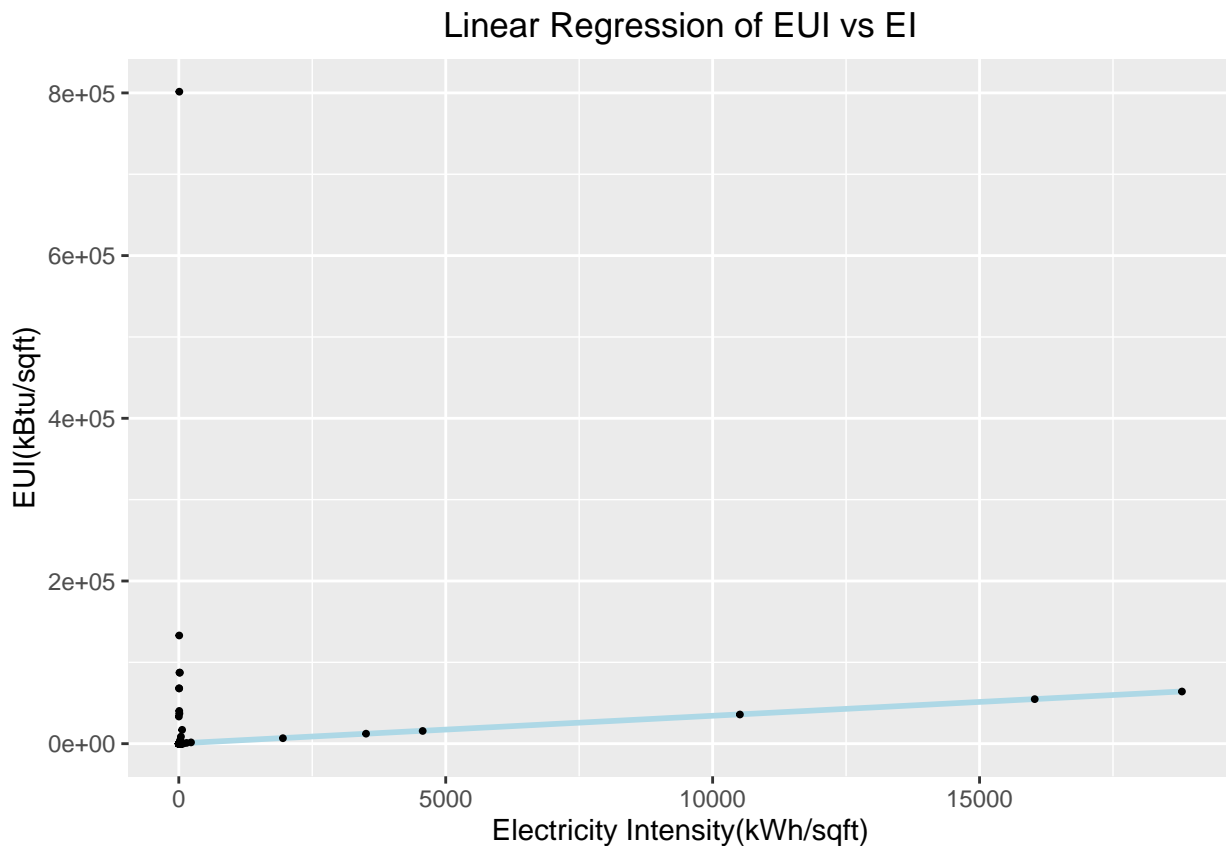
```

    Total_GHG_Emissions_Metric_Tons_CO2e,
    Municipally_Supplied_Potable_Water_Indoor_Intensity_gal_per_sqft
  ) %>%
na.omit() %>% # quit properties with missing values
rename(Record = Record_Number, # rename as the column names are too long
       EUI = Site_EUI_kBtu_per_sqft,
       EI = Weather_Normalized_Site_Electricity_Intensity_kWh_per_sqft,
       NGI = Weather_Normalized_Site_Natural_Gas_Intensity_therms_per_sqft,
       GHG = Total_GHG_Emissions_Metric_Tons_CO2e,
       WI = Municipally_Supplied_Potable_Water_Indoor_Intensity_gal_per_sqft)

# Fit in linear model with different X variables
# Create a scatterplot with regression line
ggplot(data = my_data, aes(x = EI, y = EUI)) +
  geom_smooth(method = "lm", se = FALSE, color = "lightblue") +
  geom_point(size = 0.7) +
  labs(x = "Electricity Intensity(kWh/sqft)", y = "EUI(kBtu/sqft)", title = "Linear Regression of EUI vs EI")
  theme(plot.title = element_text(hjust = 0.5)) # make title in the middle

## `geom_smooth()` using formula = 'y ~ x'

```

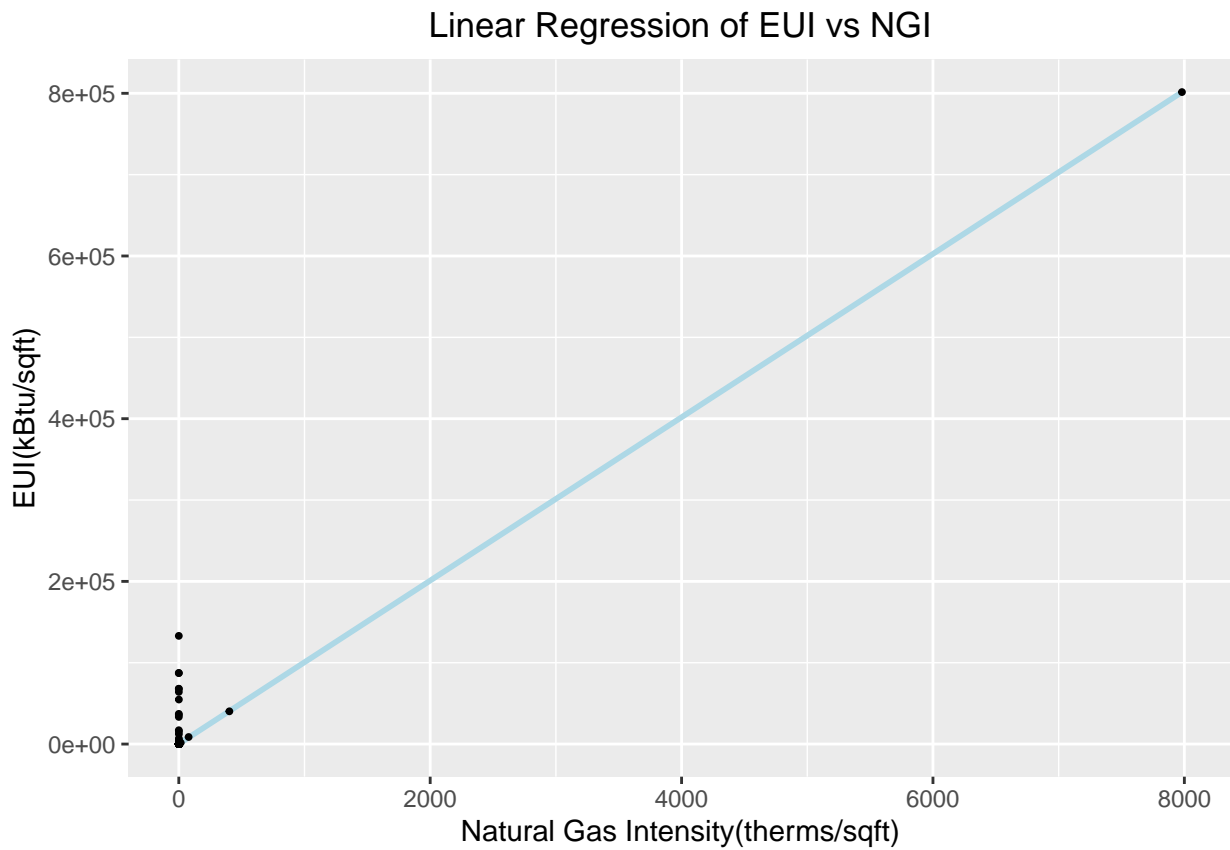


```

ggplot(data = my_data, aes(x = NGI, y = EUI)) +
  geom_smooth(method = "lm", se = FALSE, color = "lightblue") +
  geom_point(size = 0.7) +
  labs(x = "Natural Gas Intensity(therms/sqft)", y = "EUI(kBtu/sqft)", title = "Linear Regression of EUI vs NGI")
  theme(plot.title = element_text(hjust = 0.5))

```

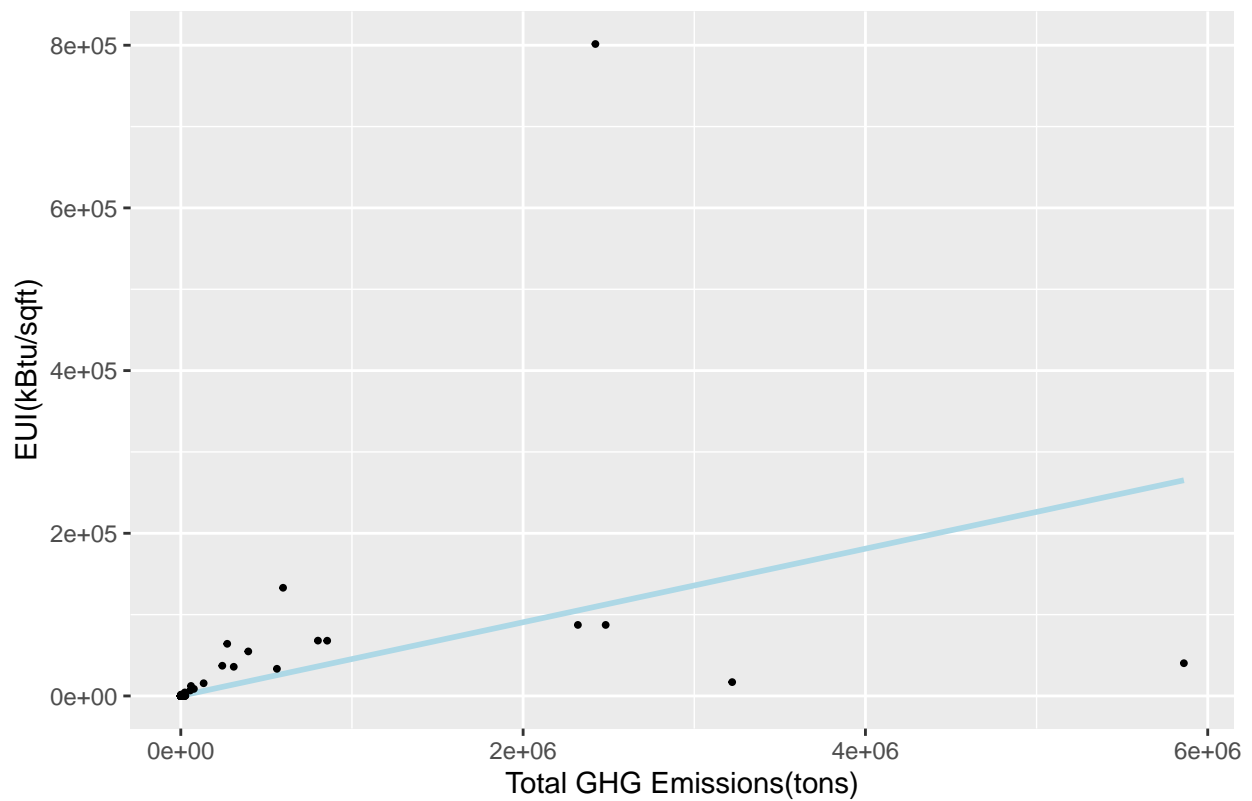
```
## `geom_smooth()` using formula = 'y ~ x'
```



```
ggplot(data = my_data, aes(x = GHG, y = EUI)) +  
  geom_smooth(method = "lm", se = FALSE, color = "lightblue") +  
  geom_point(size = 0.7) +  
  labs(x = "Total GHG Emissions(tons)", y = "EUI(kBtu/sqft)", title = "Linear Regression of EUI vs GHG") +  
  theme(plot.title = element_text(hjust = 0.5))
```

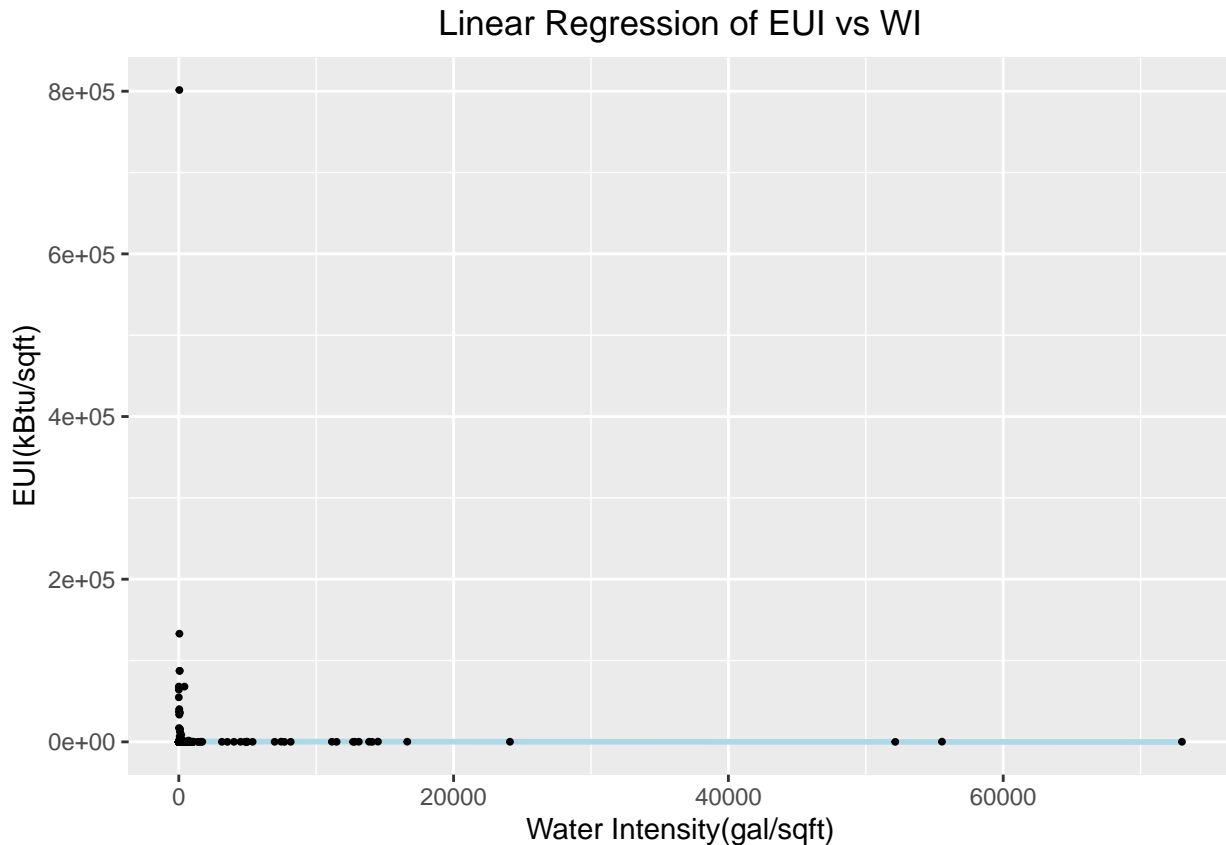
```
## `geom_smooth()` using formula = 'y ~ x'
```

Linear Regression of EUI vs GHG



```
ggplot(data = my_data, aes(x = WI, y = EUI)) +  
  geom_smooth(method = "lm", se = FALSE, color = "lightblue") +  
  geom_point(size = 0.7) +  
  labs(x = "Water Intensity(gal/sqft)", y = "EUI(kBtu/sqft)", title = "Linear Regression of EUI vs WI")  
  theme(plot.title = element_text(hjust = 0.5))
```

```
## `geom_smooth()` using formula = 'y ~ x'
```



Section #02: Spatial Join

The purpose of the spatial join operation is to join the data from American Community Survey(ACS) collected by Census Bureau to the 2015 building energy consumption benchmarking data collected under NYC Local Law 84/133 Energy Benchmarking.

Data source for spatial join

- 2016 LL33 Data Disclosure for CY2015 reporting, Government of New York City. <https://www.nyc.gov/site/buildings/codes/benchmarking.page>
- 2010 Shapefiles of NYC census blocks, United States Census Bureau. <https://www.census.gov/geographies/mapping-files/time-series/geo/tiger-line-file.html>
- 2023 Shapefiles of NYC Tax Lot (BBL), NYC Open Data. <https://www.nyc.gov/site/planning/data-maps/open-data.page#pluto>

Step #01: shapefile importing and visualization

Import the .shp files to and visualize the polygons in Arcgis Pro

Step #02: spatial join using Tax Lot data and Census block data

Spatial join with options: - Target features: Tax Lot data - Join features: Census block data - Join operation: One to one - Match option: Within - Fields to join: GeoId (the only required feature for joining ACS data)

This generate a new Tax Lot data table with a new column of the Census Block it belongs to. This means we have BBL number and Census Block geoID in each row.

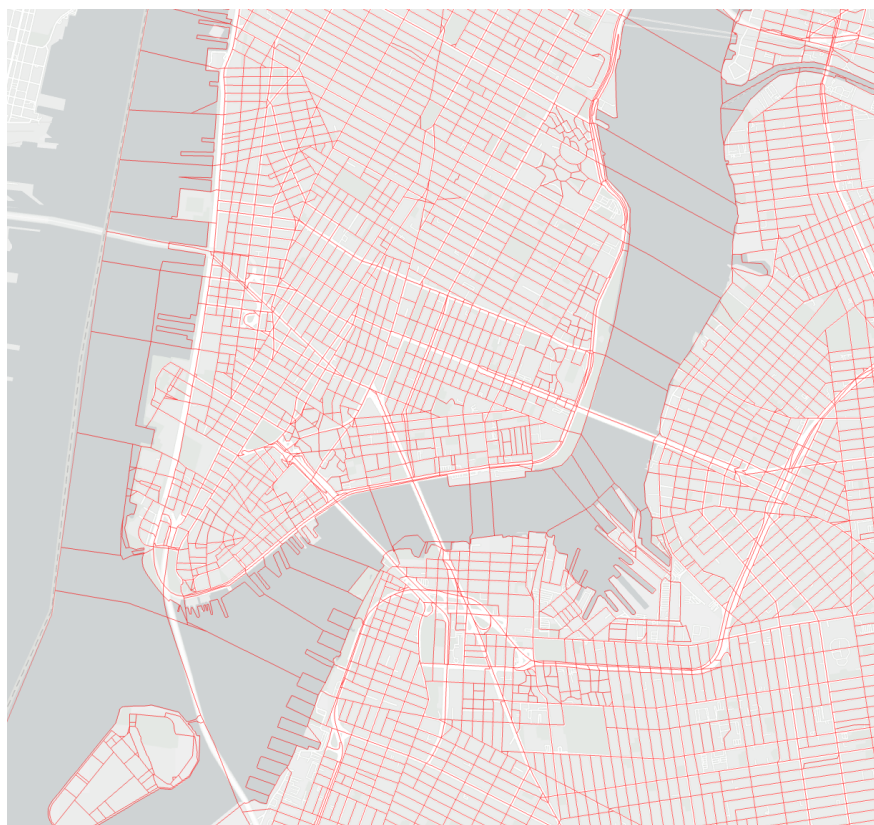


Figure 1: census block boundaries

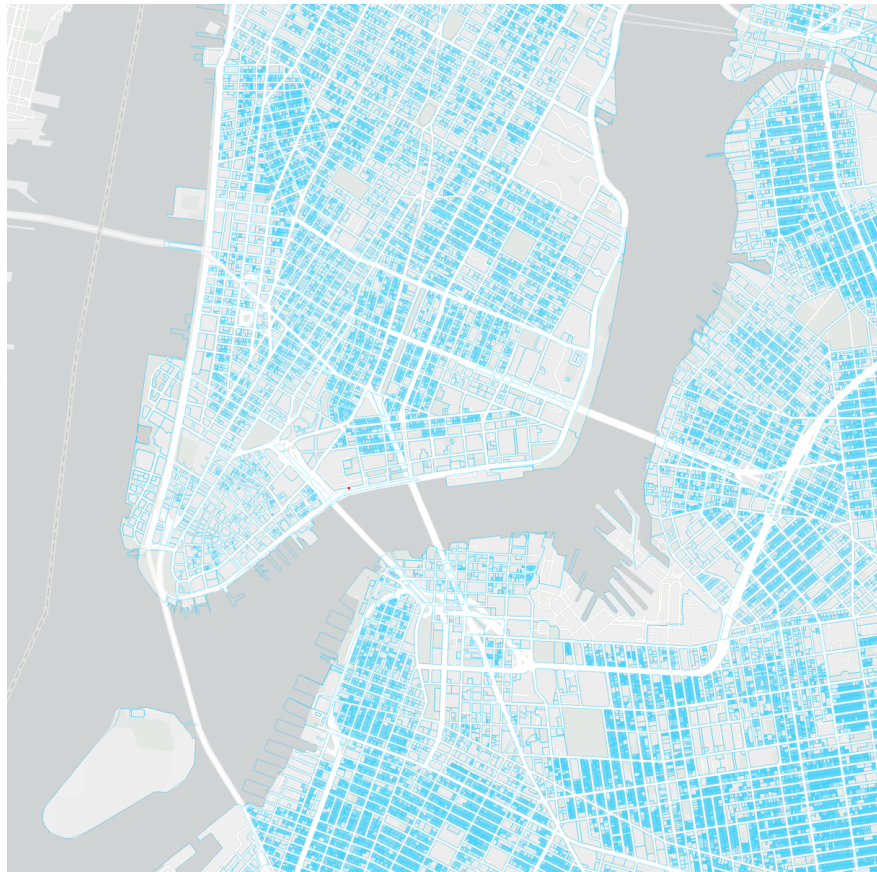


Figure 2: tax lot boundaries

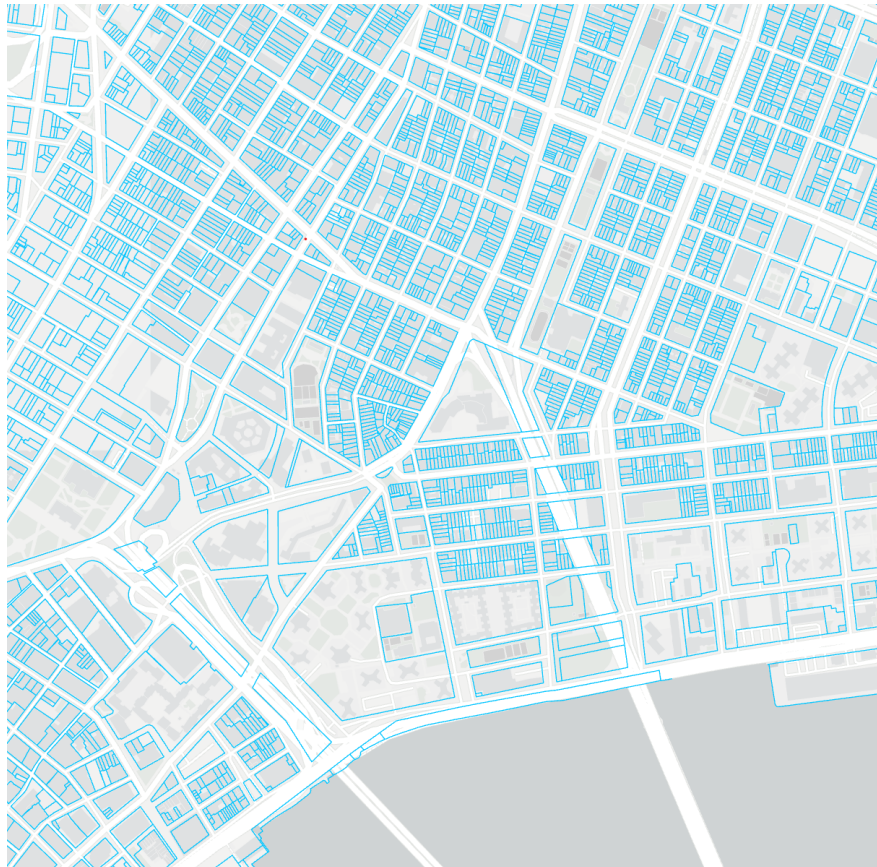


Figure 3: tax lot boundaries zoomed



Figure 4: overlay

Step #03: join building energy data

Join the geoID data to the building energy dataset using the field of BBL. This results in a new building energy use data table with a new column of Census Block geoID information it belongs to.

```
# load benchmarking table
BR <- read.csv(file = paste0(file.path(dirname(rstudioapi::getSourceEditorContext()$path)),
                             "/dataset/NYC_Building_Energy_with_GEOID/table.csv"),
               header=TRUE) %>%
  data.frame()

head(BR)
```

```
##   OID_ Record_Number Order_      BBL Coreported_BBL_Status BBLs_Coreported
## 1   NA              NA  12515  2034327503
## 2   NA          2666985   6733  2036000004
## 3   NA          2639029   6751  2036720001
## 4   NA              NA  12519  2036040001
## 5   NA          2682362   6722  2035630005
## 6   NA          4402905   5381  2023090001
##
##           Reported_BINs           Property_Name
## 1
## 2           2092718           731-755 White Plains Road
## 3           2022645           1921-1965 Lafayette Park Lane
## 4
## 5 2116665;2819754;2819749           Gold - 669 White Plains Rd
## 6           2093995  Carnegie Management: 112 Lincoln Ave
##
##           Parent_Property_Id           Parent_Property_Name
## 1
## 2 Not Applicable: Standalone Property Not Applicable: Standalone Property
## 3 Not Applicable: Standalone Property Not Applicable: Standalone Property
## 4
## 5 Not Applicable: Standalone Property Not Applicable: Standalone Property
## 6 Not Applicable: Standalone Property Not Applicable: Standalone Property
##   Street_Number      Street_Name Zip_Code Borough
## 1           329  ADMIRAL LANE         10473  Bronx
## 2          1850  LAFAYETTE AVENUE         10473  Bronx
## 3          1965  LAFAYETTE AVENUE         10473  Bronx
## 4           700  WHITE PLAINS ROAD         10473  Bronx
## 5           669  WHITE PLAINS ROAD         10473  Bronx
## 6          112  LINCOLN AVENUE         10454  Bronx
##   DOF_Benchmarking_Submission_Status Primary_Property_Type
## 1                               In Violation
## 2                In Compliance  Multifamily Housing
## 3                In Compliance  Multifamily Housing
## 4                               In Violation
## 5                In Compliance  Multifamily Housing
## 6                In Compliance  Multifamily Housing
##
##           List_of_All_Property_Use_Types_at_Property
## 1
## 2 Medical Office, Multifamily Housing, Retail Store, Supermarket/Grocery Store
## 3                                     Multifamily Housing
## 4
## 5                Multifamily Housing, Office, Parking
## 6                Multifamily Housing
```

##	Largest_Property_Use_Type	Largest_Property_Use_Type_Gross_Floor_Area_sqft
## 1		NA
## 2	Multifamily Housing	606627
## 3	Multifamily Housing	400933
## 4		NA
## 5	Multifamily Housing	50604
## 6	Multifamily Housing	90000
##	X2nd_Largest_Property_Use_Type	
## 1		
## 2	Medical Office	
## 3	Not Available	
## 4		
## 5	Parking	
## 6	Not Available	
##	X2nd_Largest_Property_Use_Type_Gross_Floor_Area_sqft	
## 1		
## 2		36180
## 3		Not Available
## 4		
## 5		8635
## 6		Not Available
##	X3rd_Largest_Property_Use_Type	
## 1		
## 2	Supermarket/Grocery Store	
## 3	Not Available	
## 4		
## 5	Office	
## 6	Not Available	
##	X3rd_Largest_Property_Use_Type_Gross_Floor_Area_sqft	Year_Built
## 1		NA
## 2		9638 1977
## 3		Not Available 1969
## 4		NA
## 5		334 2009
## 6		Not Available 1920
##	Number_of_Buildings_Self_reported	Occupancy Metered_Areas_Energy
## 1	NA	NA
## 2	0	100 Whole Building
## 3	1	100 Whole Building
## 4	NA	NA
## 5	1	100 Whole Building
## 6	1	100 Whole Building
##	Metered_Areas_Water ENERGY_STAR_Score	Site_EUI_kBtu_per_sqft
## 1	NA	NA
## 2	Not Available	94 70.9
## 3	Not Available	61 89.0
## 4	NA	NA
## 5	Not Available	NA 78.1
## 6	Not Available	74 46.2
##	Weather_Normalized_Site_EUI_kBtu_per_sqft	
## 1		NA
## 2		70.9
## 3		NA
## 4		NA

##	5		79.0	
##	6		46.1	
##		Weather_Normalized_Site_Electricity_Intensity_kWh_per_sqft		
##	1		NA	
##	2		3.1	
##	3		4.6	
##	4		NA	
##	5		7.4	
##	6		7.0	
##		Weather_Normalized_Site_Natural_Gas_Intensity_therms_per_sqft		
##	1		NA	
##	2		0.6	
##	3		0.6	
##	4		NA	
##	5		0.5	
##	6		0.2	
##		Source_EUI_kBtu_per_sqft	Weather_Normalized_Source_EUI_kBtu_per_sqft	
##	1	NA	NA	
##	2	97.6	96.5	
##	3	126.6	NA	
##	4	NA	NA	
##	5	135.7	136.0	
##	6	98.5	98.3	
##		Fuel_Oil_1_Use__kBtu	Fuel_Oil_2_Use__kBtu	Fuel_Oil_4_Use__kBtu
##	1			
##	2	Not Available	Not Available	Not Available
##	3	Not Available	4124640.6	Not Available
##	4			
##	5	Not Available	Not Available	Not Available
##	6	Not Available	Not Available	Not Available
##		Fuel_Oil_5_6_Use__kBtu	Diesel_2_Use_kBtu	District_Steam_Use_kBtu
##	1			
##	2	Not Available	Not Available	Not Available
##	3	Not Available	Not Available	Not Available
##	4			
##	5	Not Available	Not Available	Not Available
##	6	Not Available	Not Available	Not Available
##		District_Hot_Water_Use_kBtu	District_Chilled_Water_Use_kBtu	
##	1			
##	2	Not Available	Not Available	
##	3	Not Available	Not Available	
##	4			
##	5	Not Available	Not Available	
##	6	Not Available	Not Available	
##		Natural_Gas_Use_kBtu	Weather_Normalized_Site_Natural_Gas_Use_therms	
##	1	NA	NA	
##	2	39383800	396617.5	
##	3	25111336	255164.1	
##	4	NA	NA	
##	5	2669935	27288.8	
##	6	2008800	19935.2	
##		Electricity_Use_Grid_Purchase_kBtu	Weather_Normalized_Site_Electricity_kWh	
##	1	NA	NA	
##	2	7284016	2042628.6	

## 3	6438830	1845628.4
## 4	NA	NA
## 5	1308722	379132.1
## 6	2151972	630706.9
##	Total_GHG_Emissions_Metric_Tons_CO2e	Direct_GHG_Emissions_Metric_Tons_CO2e
## 1	NA	NA
## 2	2696.0	2091.9
## 3	2173.9	1639.9
## 4	NA	NA
## 5	250.4	141.8
## 6	285.2	106.7
##	Indirect_GHG_Emissions_Metric_Tons_CO2e	DOF_Property_Floor_Area_sqft
## 1	NA	110952
## 2	604.1	1021752
## 3	534.0	400932
## 4	NA	78347
## 5	108.5	58234
## 6	178.5	89275
##	Property_GFA_Self_reported_sqft	Water_Use_All_Water_Sources_kgal
## 1	NA	NA
## 2	657996	NA
## 3	400933	NA
## 4	NA	NA
## 5	50938	3821
## 6	90000	NA
##	Municipally_Supplied_Potable_Water_Indoor_Intensity_gal_per_sqft	
## 1		NA
## 2		NA
## 3		NA
## 4		NA
## 5		75.01
## 6		NA
##	Release_Date	DEP_Provided_Water_Use_kgal
## 1		NA
## 2	6/17/2016 11:59	NA
## 3	5/27/2016 10:47	NA
## 4		NA
## 5	7/31/2016 17:10	3821
## 6	4/27/2016 10:08	NA
##	Automatic_Water_Benchmarking_Eligible	Reported_Water_Method
## 1		BBL_1
## 2		2034327503
## 3		2036000004
## 4		2036720001
## 5	Eligible	2036040001
## 6		ABS 2035630005
		2023090001
##	GEOID10	Shape_Length
## 1	3.6005e+14	Shape_Area
## 2	1545.1905	125824.45
## 3	4510.0241	167236.53
## 4	1533.9754	113607.34
## 5	812.9707	41278.00
## 6	1014.5318	61452.82
## 7	846.4841	23530.14

Section #03: Next Step - Join ACS dataset to building energy dataset using the BR above