

Key Factors Driving Load Growth in PJM and ERCOT

https:

//github.com/jazpritch/KarneiKogaWargoPritchett_ENV872_EDA_FinalProject.git

Hanna Karnei, Haru Koga, Jazmine Pritchett, Jaimie Wargo

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```
# Set your ggplot theme
mytheme <- theme_gray() +
  theme(plot.title = element_text(size = 16, hjust= 0),
        axis.title = element_text(size = 13),
        legend.position = 'right')

theme_set(mytheme)
```

Load in data

```
# Read the CSV files
x2010<- read_excel("Data/Raw/ERCOT_Source_Data/2010_ercot_hourly_load_data.xls") %>%
  rename(Hour_Ending = Hour_End) %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2011<- read_excel("Data/Raw/ERCOT_Source_Data/2011_ercot_hourly_load_data.xls") %>%
  rename(Hour_Ending = Hour_End) %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2012<- read_excel("Data/Raw/ERCOT_Source_Data/2012_ercot_hourly_load_data.xls") %>%
  rename(Hour_Ending = Hour_End) %>%
```

```

    mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2013<- read_excel("Data/Raw/ERCOT_Source_Data/2013_ercot_hourly_load_data.xls") %>%
  rename(Hour_Ending = Hour_End) %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2014<- read_excel("Data/Raw/ERCOT_Source_Data/2014_ercot_hourly_load_data.xls") %>%
  rename(Hour_Ending = Hour_End) %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2015<- read_excel("Data/Raw/ERCOT_Source_Data/native_load_2015.xls") %>%
  rename(Hour_Ending = Hour_End) %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2016<- read_excel("Data/Raw/ERCOT_Source_Data/native_Load_2016.xlsx") %>%
  rename(Hour_Ending = Hour_End) %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2017<- read_excel("Data/Raw/ERCOT_Source_Data/native_Load_2017.xlsx") %>%
  rename(Hour_Ending = "Hour Ending") %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2018<- read_excel("Data/Raw/ERCOT_Source_Data/Native_Load_2018.xlsx") %>%
  rename(Hour_Ending = "HourEnding") %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2019<- read_excel("Data/Raw/ERCOT_Source_Data/Native_Load_2019.xlsx") %>%
  rename(Hour_Ending = "HourEnding") %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2020<- read_excel("Data/Raw/ERCOT_Source_Data/Native_Load_2020.xlsx") %>%
  rename(Hour_Ending = "HourEnding") %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2021<- read_excel("Data/Raw/ERCOT_Source_Data/Native_Load_2021.xlsx") %>%
  rename(Hour_Ending = "Hour Ending") %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))
x2022<- read_excel("Data/Raw/ERCOT_Source_Data/Native_Load_2022.xlsx") %>%
  rename(Hour_Ending = "Hour Ending") %>%
  mutate(Hour_Ending = as.Date(Hour_Ending, format = "%m/%d/%Y"))

# Merge the data frames based on a common column
ercot <- bind_rows(x2010, x2011, x2012, x2013, x2014, x2015, x2016, x2017, x2018, x2019, x2020, x2021, :)

ercot_load <- ercot %>%
  select(Hour_Ending,ERCOT)

```

Rationale and Research Questions

Dataset Information

Exploratory Analysis

Distribution of ERCOT hourly load data, as well as aggregated by day and month.

```
# Plot hourly data

head(ercot_load)

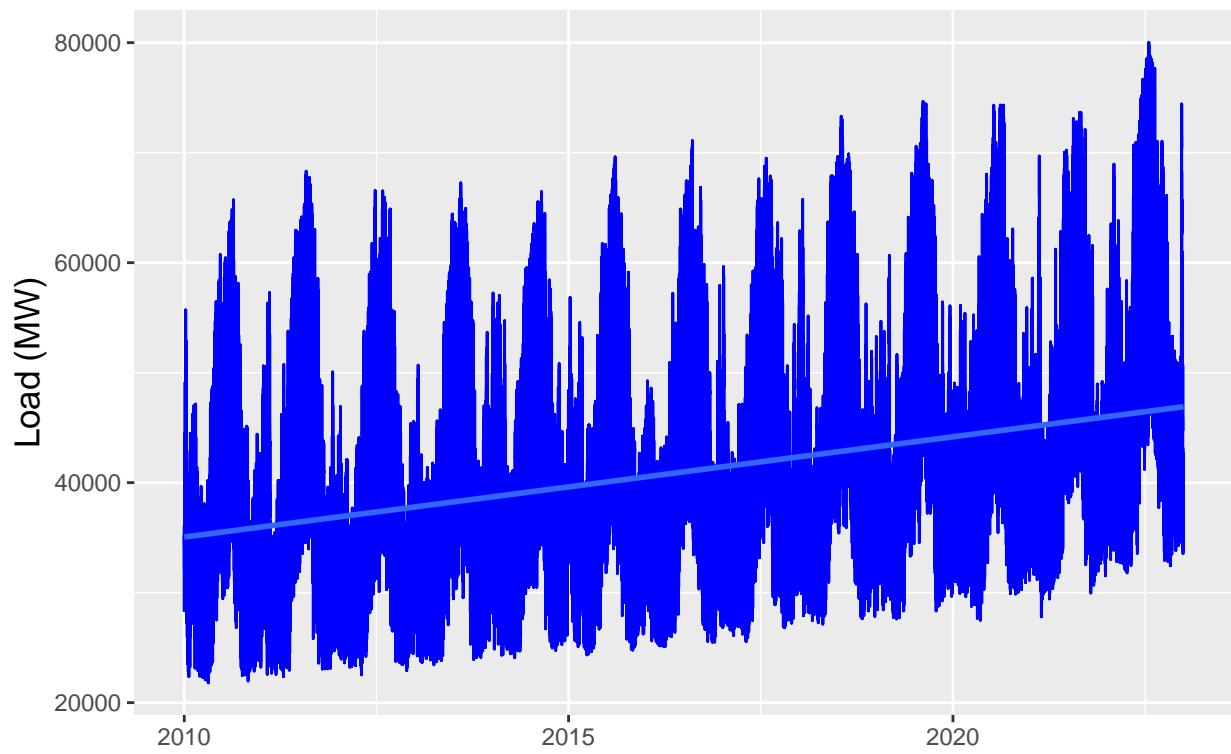
## # A tibble: 6 x 2
##   Hour_Ending   ERCOT
##   <date>        <dbl>
## 1 2010-01-01  32094.
## 2 2010-01-01  32171.
## 3 2010-01-01  32242.
## 4 2010-01-01  32459.
## 5 2010-01-01  33124.
## 6 2010-01-01  34315.

ggplot(ercot_load, aes(x = Hour_Ending, y = ERCOT)) +
  geom_line(color = "blue") +
  geom_smooth(method = "lm") +
  ggtitle('Hourly Electricity Load in ERCOT from 2010 to 2022') +
  labs(x = "", y = "Load (MW)")

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

## Warning: Removed 2 rows containing non-finite values ('stat_smooth()').
## Warning: Removed 1 row containing missing values ('geom_line()').
```

Hourly Electricity Load in ERCOT from 2010 to 2022



```
#Aggregate data by year
```

```
ercot_load_yearly <- ercot_load %>%
  mutate(
    year = year(Hour_Ending)
  ) %>%
  group_by(year) %>%
  summarise(
    Mean_Load = mean(ERCOT, na.rm = TRUE)
  ) %>%
  drop_na()
```

```
ercot_load_yearly
```

```
## # A tibble: 13 x 2
##       year Mean_Load
##   <dbl>     <dbl>
## 1  2010     36336.
## 2  2011     38127.
## 3  2012     37000.
## 4  2013     37869.
## 5  2014     38828.
## 6  2015     39669.
## 7  2016     40006.
## 8  2017     40782.
## 9  2018     42949.
```

```

## 10 2019 43818.
## 11 2020 43463.
## 12 2021 44829.
## 13 2022 49074.

#Aggregate data by month

ercot_load_monthly <- ercot_load %>%
  mutate(
    Year = year(Hour_Ending),
    Month = month(Hour_Ending)
  ) %>%
  group_by(Year, Month) %>%
  summarise(
    Mean_Load = mean(ERCOT, na.rm = TRUE)
  )

## `summarise()` has grouped output by 'Year'. You can override using the
## `.groups` argument.

ercot_load_monthly$NewDate <-
  as.Date(paste(ercot_load_monthly$Year, ercot_load_monthly$Month, "01", sep = "-"))

head(ercot_load_monthly)

## # A tibble: 6 x 4
## # Groups:   Year [1]
##       Year Month Mean_Load NewDate
##     <dbl> <dbl>    <dbl> <date>
## 1  2010     1    34884. 2010-01-01
## 2  2010     2    35240. 2010-02-01
## 3  2010     3    29550. 2010-03-01
## 4  2010     4    29856. 2010-04-01
## 5  2010     5    36547. 2010-05-01
## 6  2010     6    44000. 2010-06-01

#Aggregate data by day

ercot_load_daily <- ercot_load %>%
  mutate(
    Year = year(Hour_Ending),
    Month = month(Hour_Ending),
    Day = day(Hour_Ending)
  ) %>%
  group_by(Year, Month, Day) %>%
  summarise(
    Mean_Load = mean(ERCOT, na.rm = TRUE)
  )

## `summarise()` has grouped output by 'Year', 'Month'. You can override using the
## `.groups` argument.

```

```

ercot_load_daily$NewDate <-
  as.Date(paste(ercot_load_daily$Year, ercot_load_daily$Month, ercot_load_daily$Day, sep = "-"))

head(ercot_load_daily)

## # A tibble: 6 x 5
## # Groups:   Year, Month [1]
##   Year Month Day Mean_Load NewDate
##   <dbl> <dbl> <int>     <dbl> <date>
## 1 2010     1     1     33080. 2010-01-01
## 2 2010     1     2     33780. 2010-01-02
## 3 2010     1     3     33892. 2010-01-03
## 4 2010     1     4     38857. 2010-01-04
## 5 2010     1     5     40971. 2010-01-05
## 6 2010     1     6     39005. 2010-01-06

graph1 <- ggplot(ercot_load_monthly, aes(x = NewDate, y = Mean_Load)) +
  geom_line(color = "blue") +
  geom_smooth(method = "lm") +
  ggtitle('Monthly Electricity Load in ERCOT from 2010 to 2022') +
  labs(x = "", y = "Average Monhtly Load (MW)")

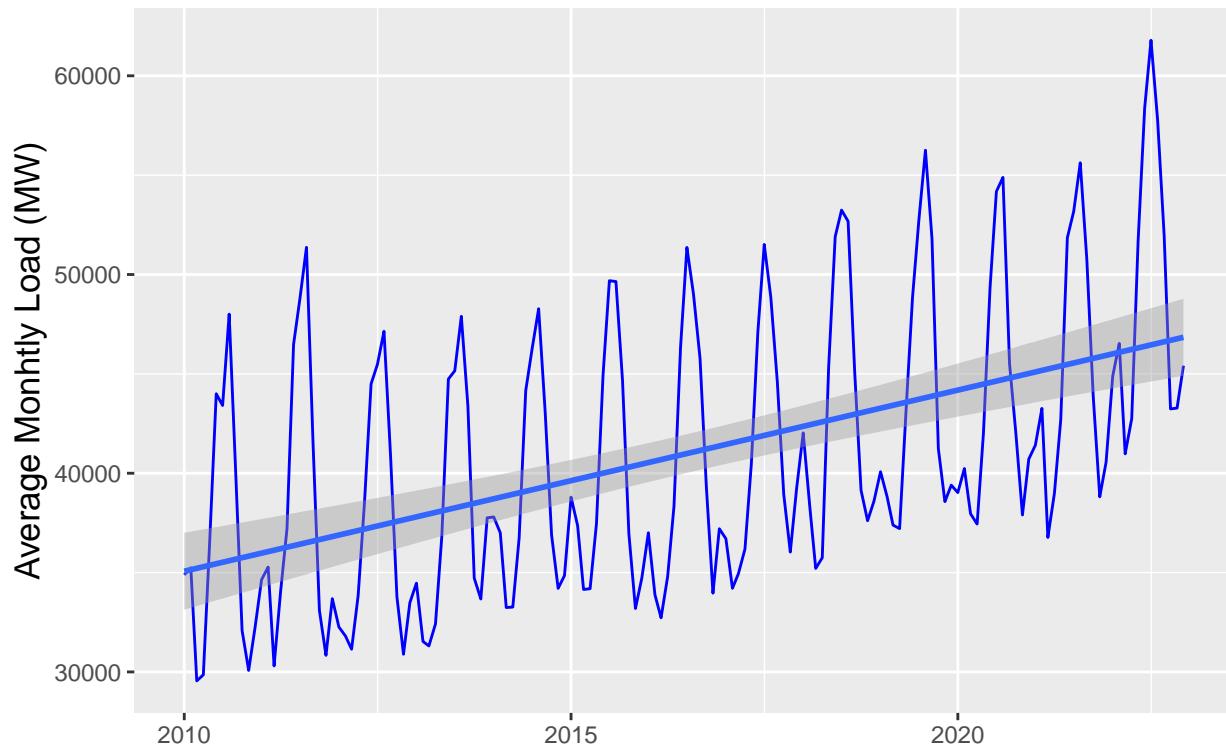
graph1

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

## Warning: Removed 1 rows containing non-finite values ('stat_smooth()').
## Warning: Removed 1 row containing missing values ('geom_line()').

```

Monthly Electricity Load in ERCOT from 2010 to 2022



```
graph2 <- ggplot(ercot_load_daily, aes(x = NewDate, y = Mean_Load)) +
  geom_line(color = "blue") +
  geom_smooth(method = "lm") +
  ggtitle('Daily Electricity Load in ERCOT from 2010 to 2022') +
  labs(x = "", y = "Average Daily Load (MW)")

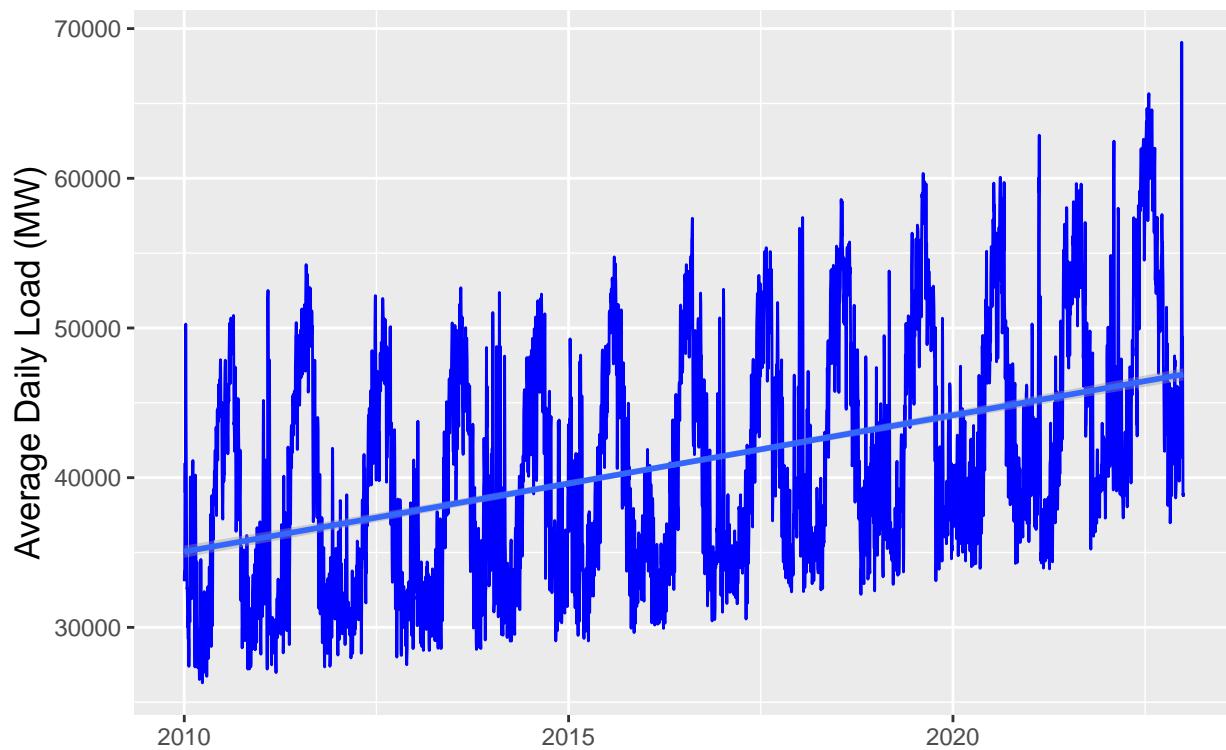
graph2

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

## Warning: Removed 1 rows containing non-finite values ('stat_smooth()').

## Warning: Removed 1 row containing missing values ('geom_line()').
```

Daily Electricity Load in ERCOT from 2010 to 2022



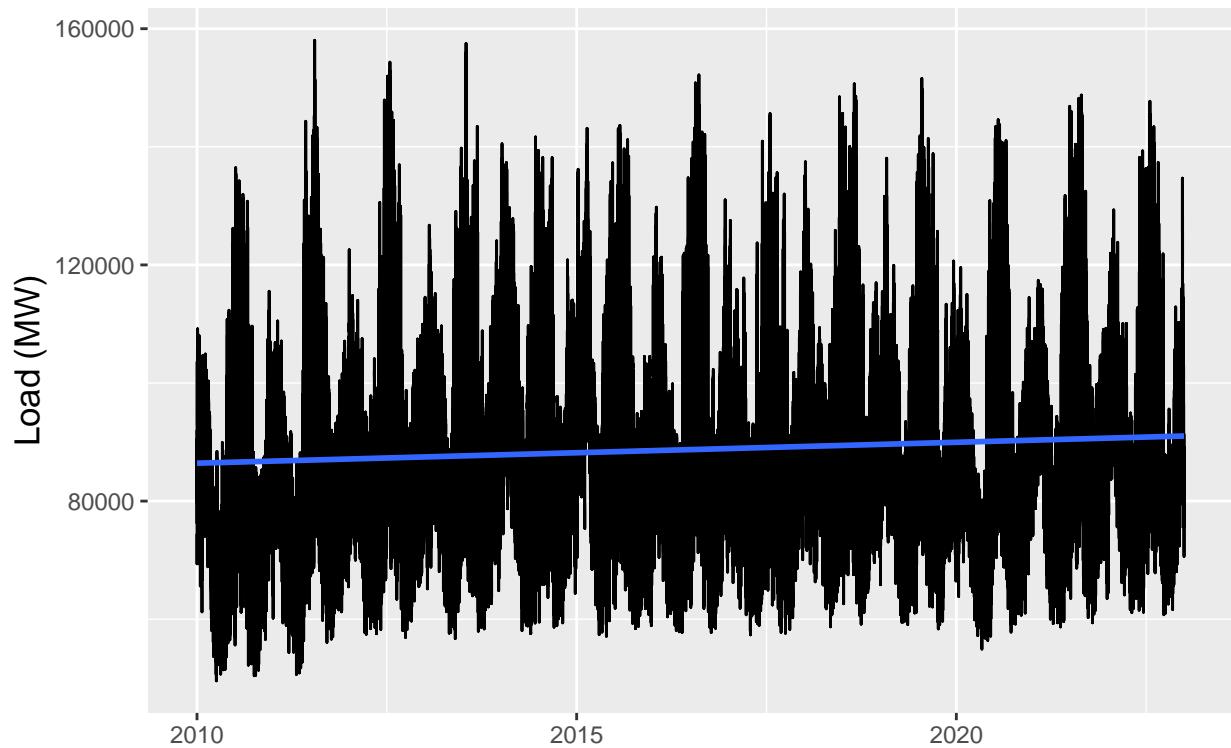
Distribution of PJM hourly load data, as well as aggregated by day and month.

```
# Plot hourly data

ggplot(PJM_load, aes(x = Date, y = Load_MW)) +
  geom_line(color = "black") +
  geom_smooth(method = "lm") +
  ggtitle('Hourly Electricity Load in PJM from 2010 to 2022') +
  labs(x = "", y = "Load (MW)")

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

Hourly Electricity Load in PJM from 2010 to 2022



```
#Aggregate data by month
```

```
PJM_load_monthly <- PJM_load %>%
  mutate(
    Year = year(Date),
    Month = month(Date)
  ) %>%
  group_by(Year, Month) %>%
  summarise(
    Mean_Load = mean(Load_MW, na.rm = TRUE)
  )
```

```
## `summarise()` has grouped output by 'Year'. You can override using the
## `.`groups` argument.
```

```
PJM_load_monthly$NewDate <-
  as.Date(paste(PJM_load_monthly$Year, PJM_load_monthly$Month, "01", sep = "-"))

head(PJM_load_monthly)
```

```
## # A tibble: 6 x 4
## # Groups:   Year [1]
##       Year Month Mean_Load NewDate
##     <dbl> <dbl>     <dbl> <date>
## 1    2010     1     88058. 2010-01-01
```

```

## 2 2010 2 86967. 2010-02-01
## 3 2010 3 74220. 2010-03-01
## 4 2010 4 67861. 2010-04-01
## 5 2010 5 73517. 2010-05-01
## 6 2010 6 88797. 2010-06-01

#Aggregate data by day

PJM_load_daily<- PJM_load %>%
  mutate(
    Year = year(Date),
    Month = month(Date),
    Day = day(Date)
  ) %>%
  group_by(Year, Month, Day) %>%
  summarise(
    Mean_Load = mean(Load_MW, na.rm = TRUE)
  )

## `summarise()` has grouped output by 'Year', 'Month'. You can override using the
## `.` argument.

PJM_load_daily$NewDate <-
  as.Date(paste(PJM_load_daily$Year, PJM_load_daily$Month,PJM_load_daily$Day, sep = "-"))

head(PJM_load_daily)

## # A tibble: 6 x 5
## # Groups:   Year, Month [1]
##   Year Month Day Mean_Load NewDate
##   <dbl> <dbl> <int>     <dbl> <date>
## 1 2010     1     1     79483. 2010-01-01
## 2 2010     1     2     89741. 2010-01-02
## 3 2010     1     3     95061. 2010-01-03
## 4 2010     1     4     98474. 2010-01-04
## 5 2010     1     5     97021. 2010-01-05
## 6 2010     1     6     95613. 2010-01-06

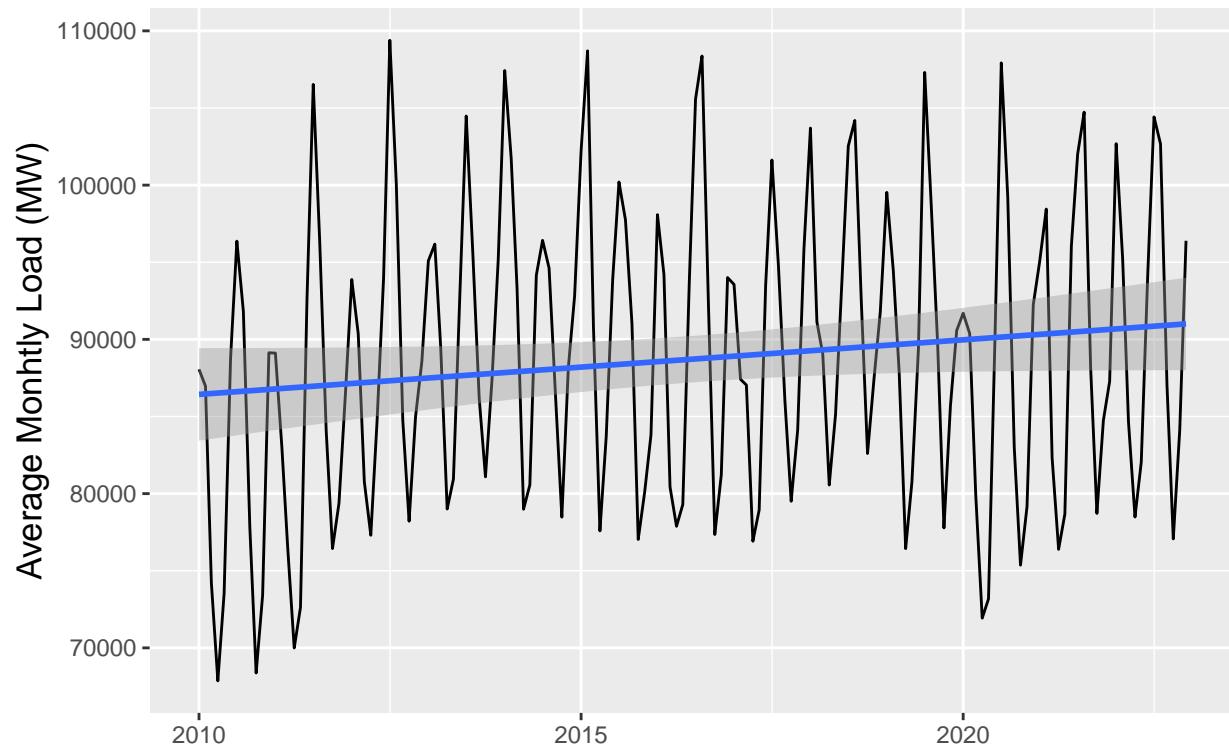
PJMgraph_month<- ggplot(PJM_load_monthly, aes(x = NewDate, y = Mean_Load)) +
  geom_line(color = "black") +
  geom_smooth(method = "lm") +
  ggtitle('Monthly Electricity Load in PJM from 2010 to 2022') +
  labs(x = "", y = "Average Monhtly Load (MW)")

PJMgraph_month

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

```

Monthly Electricity Load in PJM from 2010 to 2022

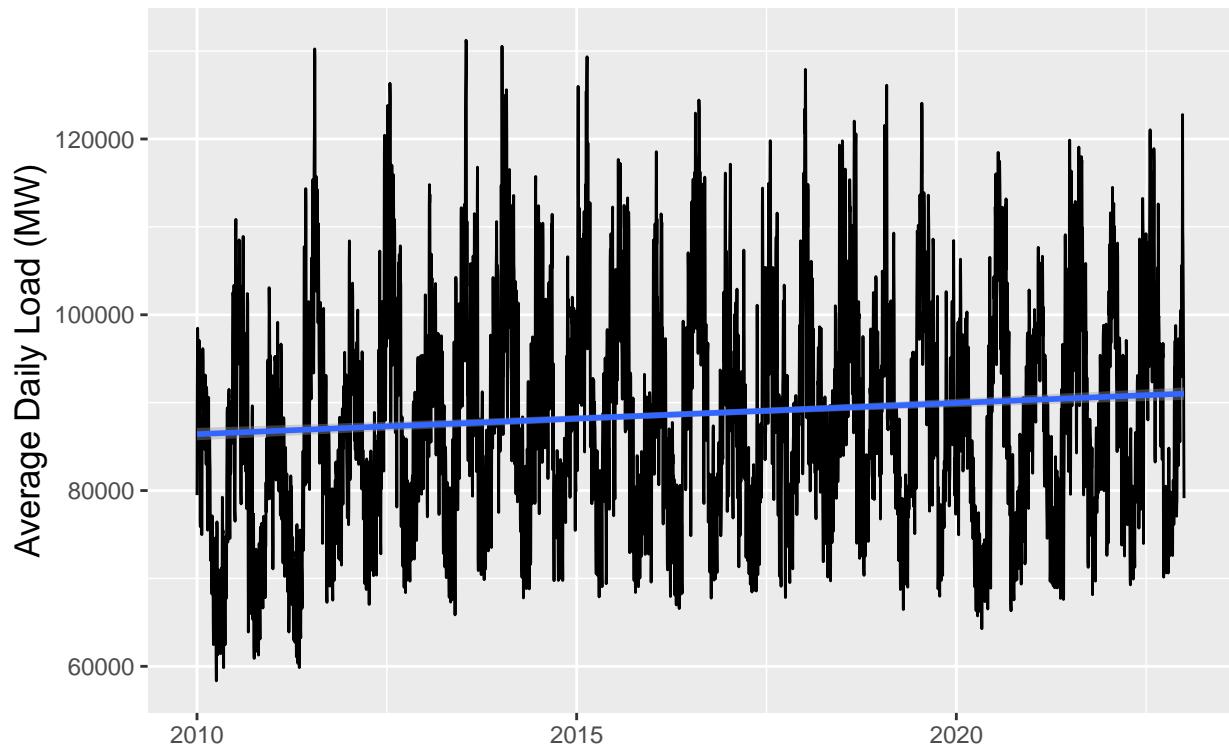


```
PJMgraph_daily <- ggplot(PJM_load_daily, aes(x = NewDate, y = Mean_Load)) +  
  geom_line(color = "black") +  
  geom_smooth(method = "lm") +  
  ggtitle('Daily Electricity Load in PJM from 2010 to 2022') +  
  labs(x = "", y = "Average Daily Load (MW)")
```

```
PJMgraph_daily
```

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

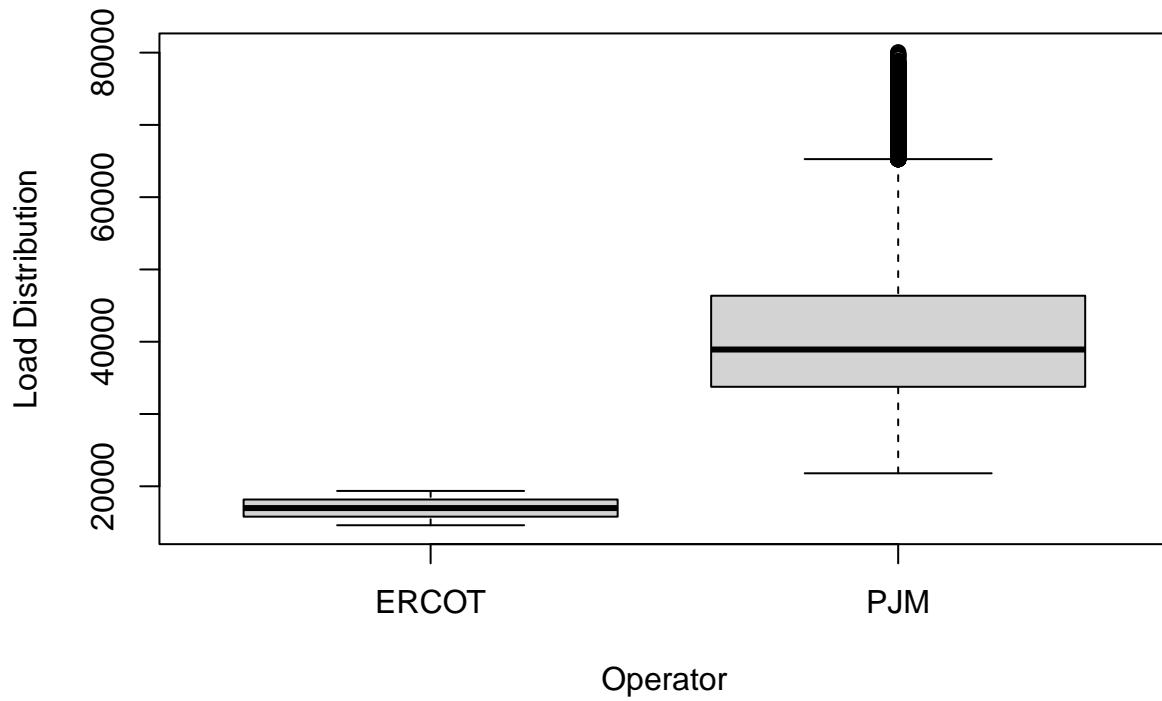
Daily Electricity Load in PJM from 2010 to 2022



Using a boxplot to compare electricity load distribution between ERCOT versus PJM.

```
boxplot(ercot_load, PJM_load, names = c("ERCOT", "PJM"), xlab = "Operator", ylab = "Load Distribution",
```

Electricity Load Distribution in ERCOT v. PJM 2010–2022



#Time Series Analysis The null hypothesis: Electricity load in ERCOT remained the same from 2010 to 2022. Alternative hypothesis: Electricity load in ERCOT has changed from 2010 to 2022.

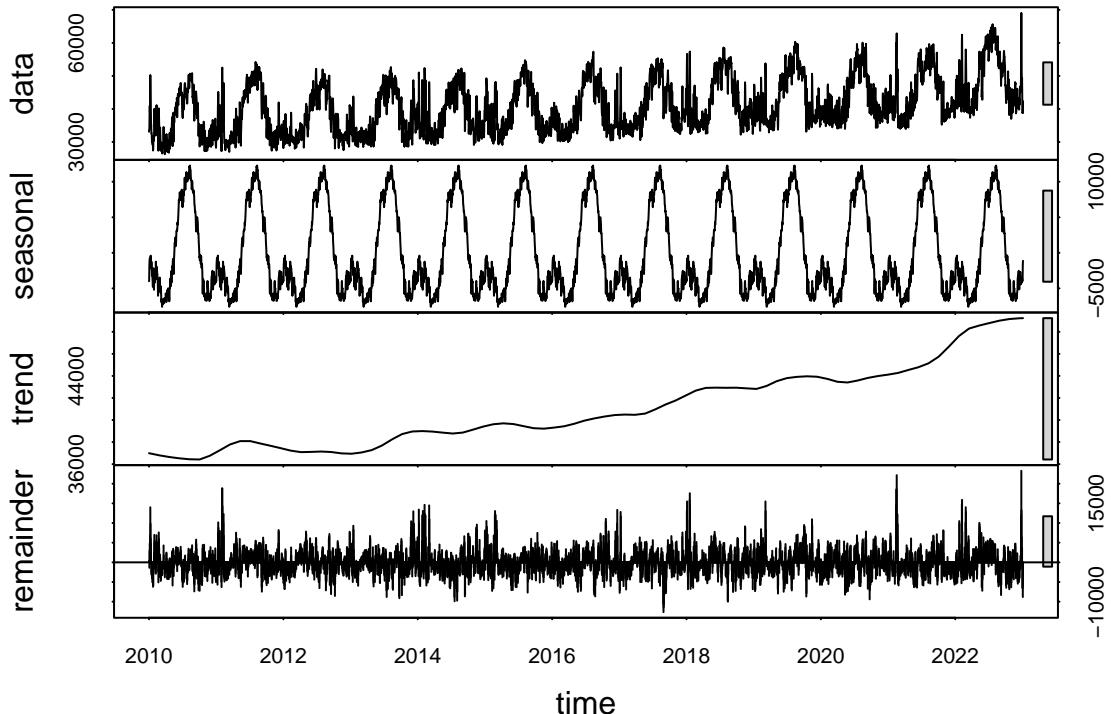
```
#Generate ts dataframes

ercot_load_daily_ts <- ts(
  ercot_load_daily$Mean_Load,
  start = c(2010,1), frequency = 365
)
```

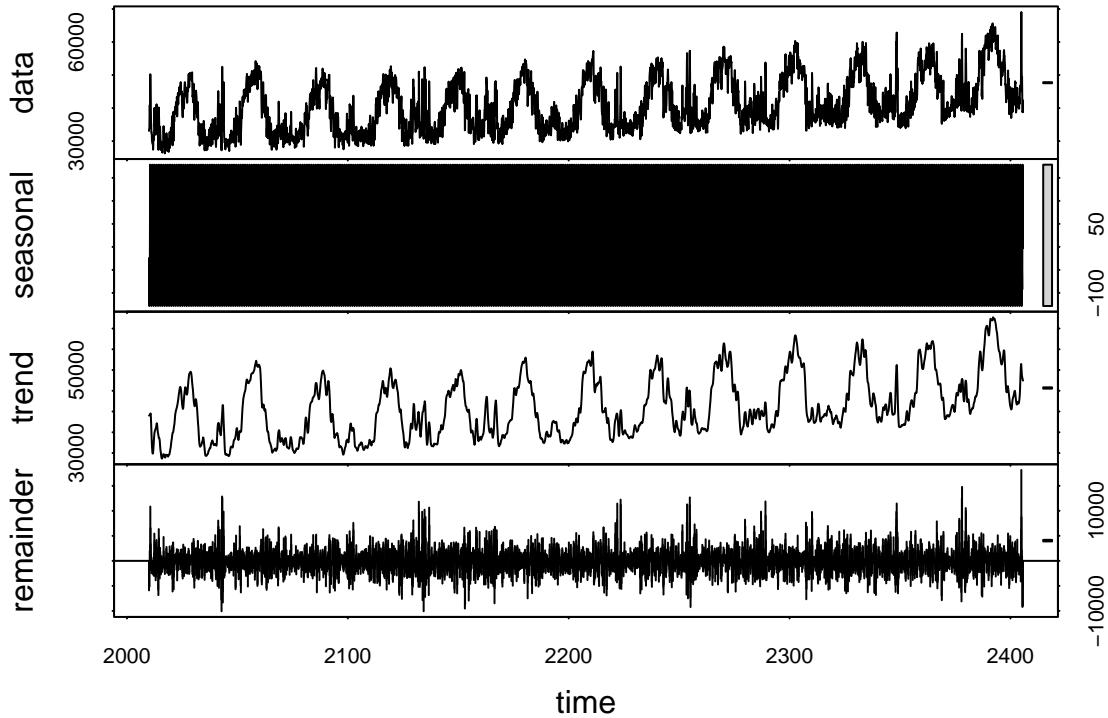
```
ercot_load_monthly_ts <- ts(
  ercot_load_daily$Mean_Load,
  start = c(2010,1),
  frequency = 12
)
```

```
#11 Decompose ts dfs and plot them
```

```
daily_decomposed <- stl(ercot_load_daily_ts, s.window="periodic")
plot(daily_decomposed)
```



```
monthly_decomposed <- stl(ercot_load_monthly_ts, s.window="periodic")
plot(monthly_decomposed)
```



```
monthly_load_trend <- Kendall::SeasonalMannKendall(ercot_load_monthly_ts)
summary(monthly_load_trend)
```

```
## Score = 287577 , Var(Score) = 82954344
## denominator = 937335.1
## tau = 0.307, 2-sided pvalue =< 2.22e-16
```

The seasonal Mann-Kendall is the most appropriate test because it is designed to analyze seasonal, non-parametric data, and there is a clear seasonal trend in our data. All other tests we learned about could only be used for non-seasonal data. A positive score (361963) and tau (0.341) indicate an increasing trend in the time series data. The p-value smaller than 0.05 allows us to reject the null hypothesis. Therefore, we can conclude that electricity demand has changed over time in ERCOT.

```
# Extract seasonal component from the time series

monthly_decomposed_components <- as.data.frame(monthly_decomposed$time.series[,1:3])

monthly_decomposed_deseasoned <-
  ercot_load_monthly_ts-monthly_decomposed_components$seasonal

# Run the Mann Kendall test on the non-seasonal data

monthly_load_trend_deseason <- Kendall::MannKendall(monthly_decomposed_deseasoned)
summary(monthly_load_trend_deseason)
```

```
## Score = 3444826 , Var(Score) = 11904225280
## denominator = 11274125
## tau = 0.306, 2-sided pvalue =< 2.22e-16
```

Based on the results of the Mann Kendall test, we are able to reject the null hypothesis ($p<0.05$) and identify an increasing trend in the data ($\tau=0.34$, $\text{score}=4339158$). Hence, the load in ERCOT has indeed changed over time; it was not simply a seasonal change. A p-value <0.05 in the seasonal Mann Kendall test also led us to the same conclusion.

The null hypothesis: Electricity load in PJM remained the same from 2010 to 2022. Alternative hypothesis: Electricity load in PJM has changed from 2010 to 2022.

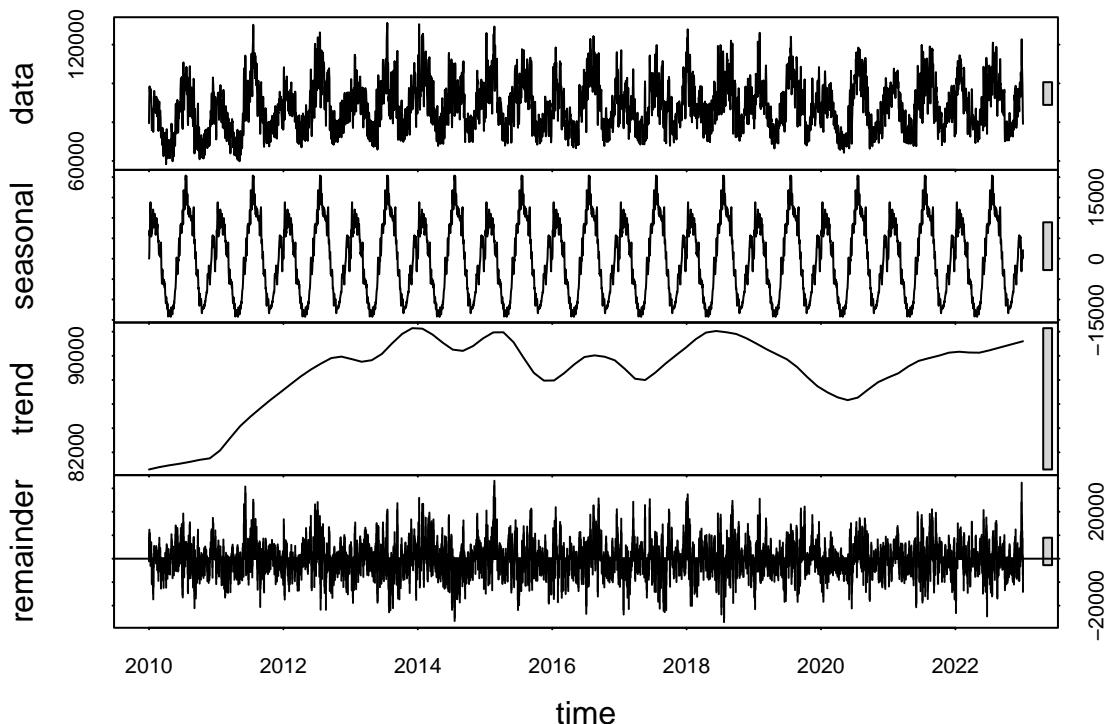
```
#Generate ts dataframes

PJM_load_daily_ts <- ts(
  PJM_load_daily$Mean_Load,
  start = c(2010,1), frequency = 365
)
```

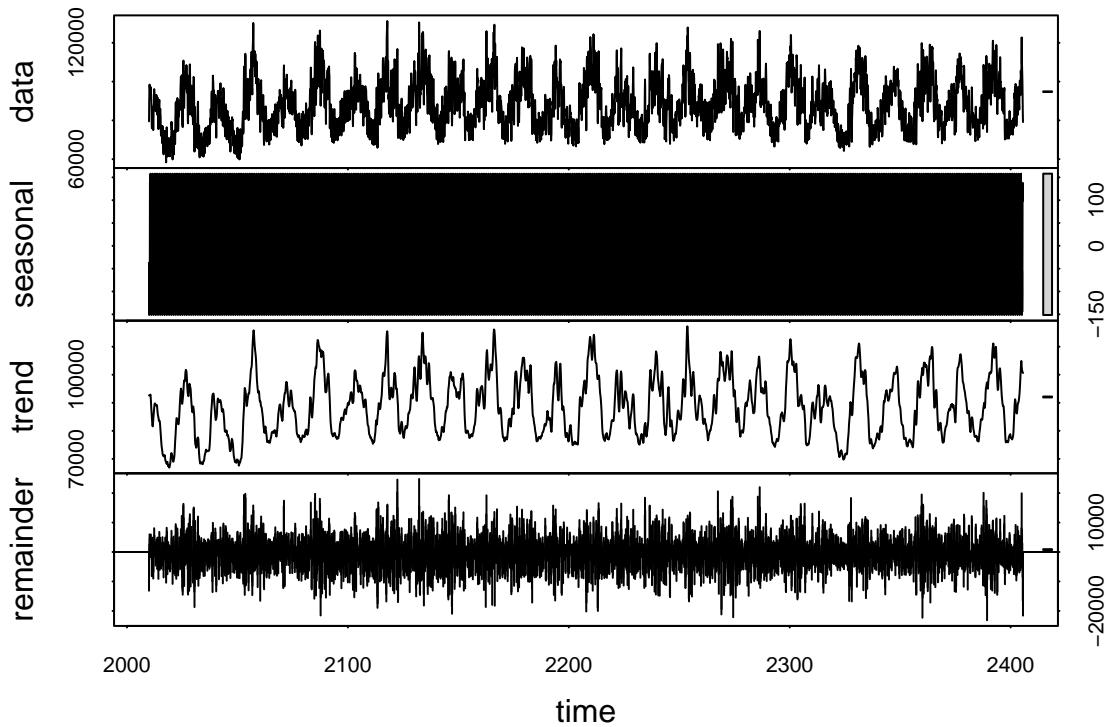
```
PJM_load_monthly_ts <- ts(
  PJM_load_daily$Mean_Load,
  start = c(2010,1),
  frequency = 12
)
```

```
#11 Decompose ts dfs and plot them
```

```
daily_decomposed_PJM <- stl(PJM_load_daily_ts, s.window="periodic")
plot(daily_decomposed_PJM)
```



```
monthly_decomposed_PJM <- stl(PJM_load_monthly_ts, s.window="periodic")
plot(monthly_decomposed_PJM)
```



```
monthly_load_trend_PJM <- Kendall::SeasonalMannKendall(PJM_load_monthly_ts)
summary(monthly_load_trend_PJM)
```

```
## Score = 63799 , Var(Score) = 82902071
## denominator = 936939.6
## tau = 0.0681, 2-sided pvalue = 2.4352e-12
```

The seasonal Mann-Kendall is the most appropriate test because it is designed to analyze seasonal, non-parametric data, and there is a clear seasonal trend in our data. All other tests we learned about could only be used for non-seasonal data. A positive score (82902071) and tau (0.0681) indicate an increasing trend in the time series data. The p-value smaller than 0.05 allows us to reject the null hypothesis. Therefore, we can conclude that electricity demand has changed over time in PJM.

```
# Extract seasonal component from the time series

monthly_decomposed_components_PJM <- as.data.frame(monthly_decomposed_PJM$time.series[,1:3])

monthly_decomposed_deseasoned_PJM <-
PJM_load_monthly_ts-monthly_decomposed_components_PJM$seasonal

# Run the Mann Kendall test on the non-seasonal data

monthly_load_trend_deseason_PJM <- Kendall::MannKendall(monthly_decomposed_deseasoned_PJM)
summary(monthly_load_trend_deseason_PJM)
```

```

## Score = 774030 , Var(Score) = 11896706048
## denominator = 11269379
## tau = 0.0687, 2-sided pvalue =< 2.22e-16

```

Mann Kendall test rejects the null hypothesis, which indicates that there is an increasing trend in the data ($\tau=0.0687$, score=11896706048). This means that PJM load has changed over time from 2010 to 2022 without seasonality components. A p-value <0.05 in the seasonal Mann Kendall test also led us to the same conclusion.

```

growth_PJM <- diff(PJM_load_monthly_ts) / stats::lag(PJM_load_monthly_ts)
#growth_ERCOT <- diff(ERCOT_load_monthly_ts) / stats::lag(ERCOT_load_monthly_ts)

#plot(Date[-1], growth_PJM, type = "l", col = "blue", ylab = "Growth Rate", xlab = "Time")
#lines(Date[-1], growth_ERCOT, type = "l", col = "red")

```

#Regression Analysis ## PJM Simple Linear Regression

Simple Linear Regression

The null hypothesis: There is no change in electricity load with price in PJM from years...

Alternative hypothesis: Electricity load in PJM has changed in relation to price from years...

Real-time pricing reflects the actual market conditions, while day-ahead pricing provides insight into the expected future market conditions. Both pricing data used to understand regression analysis. Market Performance Clearing Price (\$/MWh) used reflecting the cost of the movement of load in response to regulation signals, which is related to the performance of the resource in meeting the demand.

```

# Read the CSV files
rt_2012<- read_csv("Data/Raw/PJM_Source_Data/rt_2012.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmccp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 2209 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (7): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
## lgl (2): total_pjm_rmccp_cr, total_pjm_rmpcp_cr
##
## 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.

rt_2013<- read_csv("Data/Raw/PJM_Source_Data/rt_2013.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (7): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
## lgl (2): total_pjm_rmccp_cr, total_pjm_rmpcp_cr
##
## 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.

```

```

rt_2014<- read_csv("Data/Raw/PJM_Source_Data/rt_2014.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (7): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
## lgl (2): total_pjm_rmccp_cr, total_pjm_rmpcp_cr
##
## 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.

rt_2015<- read_csv("Data/Raw/PJM_Source_Data/rt_2015.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (7): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
## lgl (2): total_pjm_rmccp_cr, total_pjm_rmpcp_cr
##
## 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.

rt_2016<- read_csv("Data/Raw/PJM_Source_Data/rt_2016.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8783 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (7): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
## lgl (2): total_pjm_rmccp_cr, total_pjm_rmpcp_cr
##
## 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.

rt_2017<- read_csv("Data/Raw/PJM_Source_Data/rt_2017.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (7): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...

```

```

## lgl (2): total_pjm_rmccp_cr, total_pjm_rmpcp_cr
##
## 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.

rt_2018<- read_csv("Data/Raw/PJM_Source_Data/rt_2018.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (9): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
##
## 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.

rt_2019<- read_csv("Data/Raw/PJM_Source_Data/rt_2019.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (9): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
##
## 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.

rt_2020<- read_csv("Data/Raw/PJM_Source_Data/rt_2020.csv")%>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8783 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_endin...
## dbl (9): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
##
## 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.

rt_2021<- read_csv("Data/Raw/PJM_Source_Data/rt_2021.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 13
## -- Column specification -----
## Delimiter: ","

```

```

## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (9): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
##
## 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.

rt_2022<- read_csv("Data/Raw/PJM_Source_Data/rt_2022.csv") %>%
  rename(Date = datetime_beginning_ept, Price = rmpcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 13
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (9): rmccp, rmpcp, total_pjm_rt_load_mwh, total_pjm_loc_credit, total_pj...
##
## 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.

#Merge the data frames based on a common column
rt_load <- bind_rows(rt_2012, rt_2013, rt_2014, rt_2015, rt_2016, rt_2017, rt_2018, rt_2019, rt_2020, rt_2022)

# Read the CSV files
dasr_2010<- read_csv("Data/Raw/PJM_Source_Data/dasr_2010.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2011<- read_csv("Data/Raw/PJM_Source_Data/dasr_2011.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2012<- read_csv("Data/Raw/PJM_Source_Data/dasr_2012.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

```

```

## Rows: 8783 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2013<- read_csv("Data/Raw/PJM_Source_Data/dasr_2013.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2014<- read_csv("Data/Raw/PJM_Source_Data/dasr_2014.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2015<- read_csv("Data/Raw/PJM_Source_Data/dasr_2015.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2016<- read_csv("Data/Raw/PJM_Source_Data/dasr_2016.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

```

```

## Rows: 8783 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2017<- read_csv("Data/Raw/PJM_Source_Data/dasr_2017.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2018<- read_csv("Data/Raw/PJM_Source_Data/dasr_2018.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2019<- read_csv("Data/Raw/PJM_Source_Data/dasr_2019.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2020<- read_csv("Data/Raw/PJM_Source_Data/dasr_2020.csv")%>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

```

```

## Rows: 8783 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2021<- read_csv("Data/Raw/PJM_Source_Data/dasr_2021.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 8759 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

dasr_2022<- read_csv("Data/Raw/PJM_Source_Data/dasr_2022.csv") %>%
  rename(Date = datetime_beginning_ept, Price = dasrmcp, Load_MW = total_pjm_rt_load_mwh) %>%
  mutate(Date = as.Date(Date, format = "%m/%d/%Y"))

## Rows: 6550 Columns: 9
## -- Column specification -----
## Delimiter: ","
## chr (4): datetime_beginning_utc, datetime_beginning_ept, datetime_ending_utc...
## dbl (5): dasrmcp, total_pjm_rt_load_mwh, total_pjm_cleared_dasr_mwh, total_p...
##
## 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.

#Merge the data frames based on a common column
dasr_load <- bind_rows(dasr_2010, dasr_2011, dasr_2012, dasr_2013, dasr_2014, dasr_2015, dasr_2016, dasr_2017, dasr_2018, dasr_2019, dasr_2020, dasr_2021, dasr_2022)

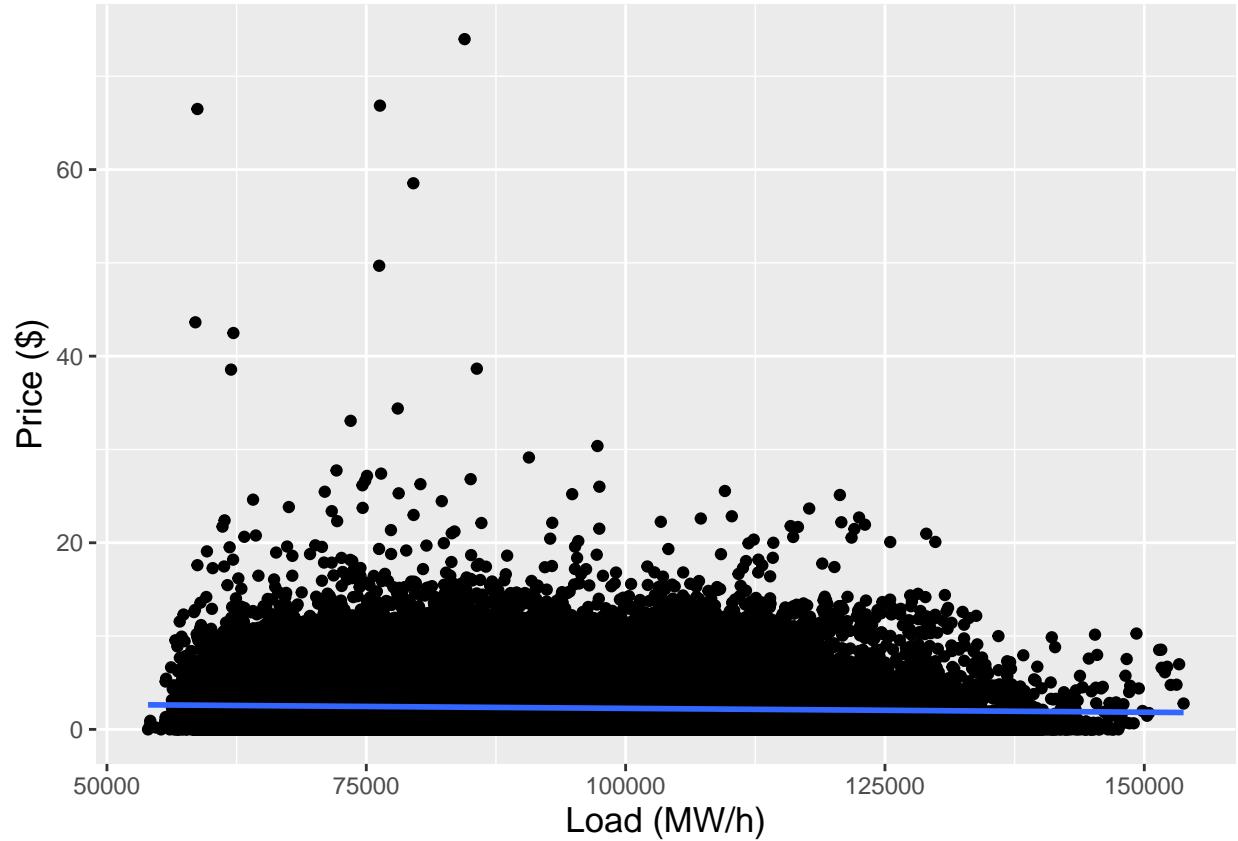
ggplot(rt_load, aes(x= Load_MW, y= Price)) +
  geom_point() +
  geom_smooth(method = "lm", se = FALSE) +
  labs(x= "Load (MW/h)", y = "Price ($)")

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

## Warning: Removed 8760 rows containing non-finite values ('stat_smooth()').

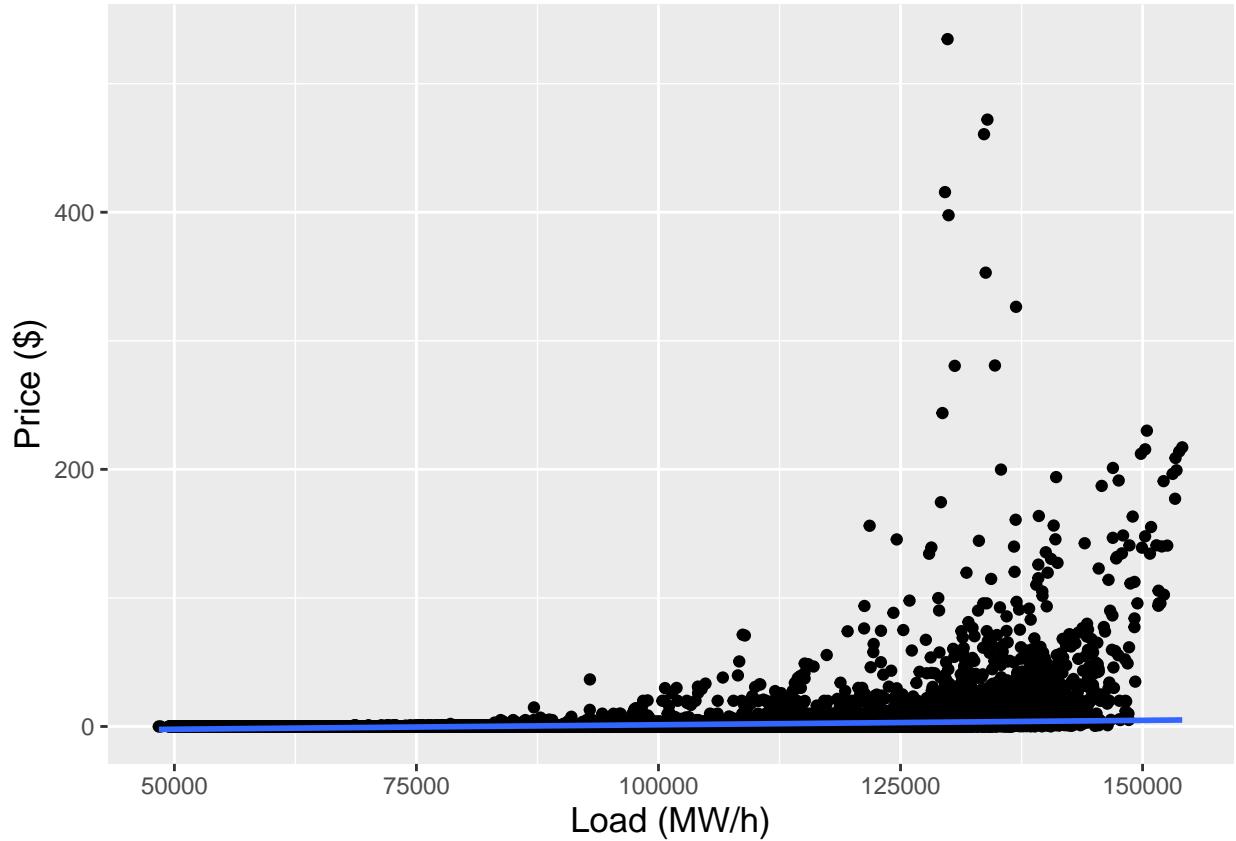
## Warning: Removed 8760 rows containing missing values ('geom_point()').

```



```
ggplot(dasr_load, aes(x= Load_MW, y= Price)) +  
  geom_point() + geom_smooth(method = "lm", se = FALSE) +  
  labs(x= "Load (MW/h)", y = "Price ($)")
```

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



```
rt.regression <- lm(data = rt_load, Price ~ Load_MW)
summary(rt.regression)
```

```
##
## Call:
## lm(formula = Price ~ Load_MW, data = rt_load)
##
## Residuals:
##      Min       1Q   Median       3Q      Max 
## -2.632 -1.810 -0.975  1.017 71.599 
## 
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 3.078e+00  5.320e-02   57.85  <2e-16 ***
## Load_MW     -8.254e-06  5.960e-07  -13.85  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 
## Residual standard error: 2.711 on 81086 degrees of freedom
## (8760 observations deleted due to missingness)
## Multiple R-squared:  0.002359, Adjusted R-squared:  0.002347 
## F-statistic: 191.8 on 1 and 81086 DF,  p-value: < 2.2e-16
```

```
cor.test(rt_load$Price, rt_load$Load_MW)
```

```

## 
## Pearson's product-moment correlation
## 
## data: rt_load$Price and rt_load$Load_MW
## t = -13.848, df = 81086, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.05543753 -0.04170422
## sample estimates:
##          cor
## -0.04857317

dasr.regression <- lm(data = dasr_load, Price ~ Load_MW)
summary(dasr.regression)

## 
## Call:
## lm(formula = Price ~ Load_MW, data = dasr_load)
## 
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.07  -0.95  -0.21   0.45 531.28
## 
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) -5.607e+00  9.390e-02 -59.71   <2e-16 ***
## Load_MW      6.921e-05  1.061e-06  65.22   <2e-16 ***
## ---      
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 
## Residual standard error: 5.777 on 111728 degrees of freedom
## Multiple R-squared:  0.03668,    Adjusted R-squared:  0.03667 
## F-statistic:  4254 on 1 and 111728 DF,  p-value: < 2.2e-16

cor.test(dasr_load$Price, dasr_load$Load_MW)

## 
## Pearson's product-moment correlation
## 
## data: dasr_load$Price and dasr_load$Load_MW
## t = 65.224, df = 111728, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.1858631 0.1971601
## sample estimates:
##          cor
## 0.191518

```

ERCOT Multiple Linear Regression

```

ercot_linear_model_data <- read.csv("Data/Raw/ERCOT_Source_Data/ERCOTLinearModelData.csv")
ercot_linear_model_data <- left_join(ercot_linear_model_data, ercot_load_yearly) %>% drop_na()

## Joining with 'by = join_by(year)'

ercot_linear_model_data$percent_renew <- stringr::str_extract(ercot_linear_model_data$percent_renew, "^\d+.\d+")
ercot_linear_model_data$percent_minority <- ercot_linear_model_data$percent_minority %>%
  as.numeric()

ercot_linear_model_data$population <- as.numeric(gsub(",","",ercot_linear_model_data$population))

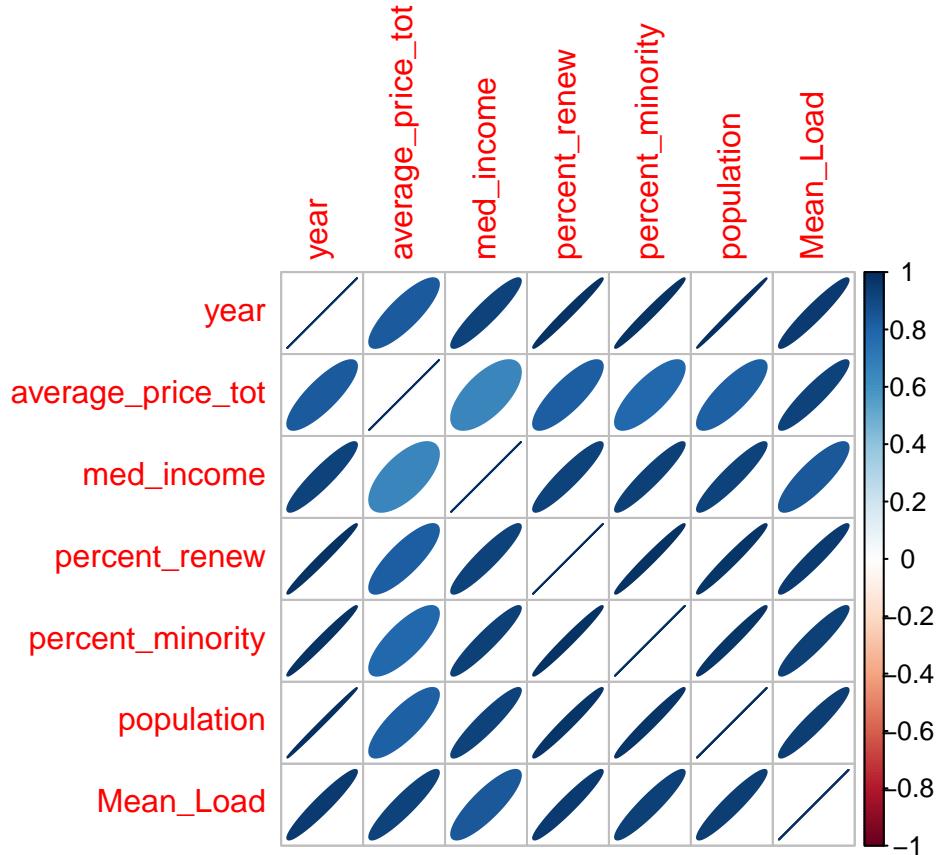
ercot_linear_model_data

##      year average_price_res average_price_com average_price_ind average_price_tot
## 1  2010        11.54          10.19           6.77         9.83
## 2  2011        11.72          10.24           6.82         9.90
## 3  2012        11.88          10.09           6.67         9.84
## 4  2013        12.13          10.26           6.89        10.07
## 5  2014        12.52          10.74           7.10        10.44
## 6  2015        12.65          10.64           6.91        10.41
## 7  2016        12.55          10.43           6.76        10.27
## 8  2017        12.89          10.66           6.88        10.48
## 9  2018        12.87          10.67           6.92        10.53
## 10 2019        13.01          10.68           6.81        10.54
## 11 2020        13.15          10.59           6.67        10.59
## 12 2021        13.66          11.22           7.18        11.10
## 13 2022        15.04          12.41           8.32        12.36
##      med_income percent_renew percent_minority population Mean_Load
## 1       61680        7.39          54.7    25241897   36335.65
## 2       62080        7.63          55.1    25645504   38127.12
## 3       64480        8.74          55.6    26084120   37000.04
## 4       63070        8.75          56.0    26479646   37869.12
## 5       65170       10.10          56.0    26963092   38828.39
## 6       68360       12.50          57.0    27468531   39669.23
## 7       69770       14.50          57.0    27914064   40006.36
## 8       70860       15.80          58.0    28291024   40781.92
## 9       69100       17.40          58.4    28624564   42949.25
## 10      76820       20.10          58.8    28986794   43817.93
## 11      77110       22.10          60.3    29232474   43463.07
## 12      72680       23.90          59.7    29558864   44828.81
## 13      74640       25.00          60.2    30029572   49074.36

ercot_linear_model_data_totprice <- ercot_linear_model_data %>%
  select(-c(average_price_com, average_price_ind, average_price_res))

ercot.tot.Corr <- cor(ercot_linear_model_data_totprice)
corrplot(ercot.tot.Corr, method = 'ellipse')

```



```

ercot.AIC.total <- lm(data = ercot_linear_model_data_totprice,
                        Mean_Load ~ average_price_tot + population + percent_renew + med_income +
                        percent_minority)

step(ercot.AIC.total)

## Start: AIC=177.64
## Mean_Load ~ average_price_tot + population + percent_renew +
##   med_income + percent_minority
##
##           Df Sum of Sq      RSS      AIC
## - percent_minority  1     961  4442811 175.64
## - population       1    4576  4446425 175.66
## - med_income        1    7472  4449321 175.66
## <none>                  4441849 177.64
## - percent_renew     1  1243905  5685754 178.85
## - average_price_tot 1  6162527 10604376 186.95
##
## Step: AIC=175.64
## Mean_Load ~ average_price_tot + population + percent_renew +
##   med_income
##
##           Df Sum of Sq      RSS      AIC
## - population       1     3624  4446435 173.66
## - med_income        1     6798  4449609 173.66

```

```

## <none>                               4442811 175.64
## - percent_renew      1   1911430  6354241 178.30
## - average_price_tot  1   6441468 10884279 185.29
##
## Step: AIC=173.65
## Mean_Load ~ average_price_tot + percent_renew + med_income
##
##          Df Sum of Sq      RSS      AIC
## - med_income      1     13525  4459959 171.69
## <none>                      4446435 173.66
## - percent_renew    1   3236589  7683024 178.76
## - average_price_tot 1   7110458 11556893 184.07
##
## Step: AIC=171.69
## Mean_Load ~ average_price_tot + percent_renew
##
##          Df Sum of Sq      RSS      AIC
## <none>                      4459959 171.69
## - average_price_tot  1   9019288 13479248 184.07
## - percent_renew       1  19338242 23798201 191.46

##
## Call:
## lm(formula = Mean_Load ~ average_price_tot + percent_renew, data = ercot_linear_model_data_totprice)
##
## Coefficients:
##             (Intercept)  average_price_tot      percent_renew
##               11580.9           2300.1              353.6

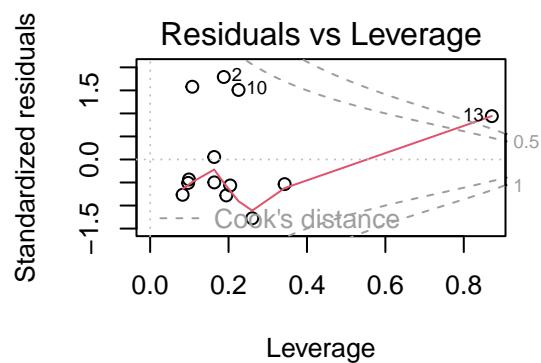
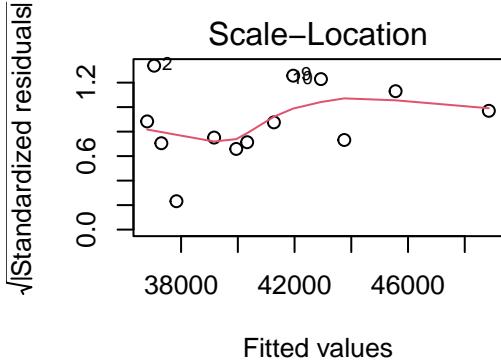
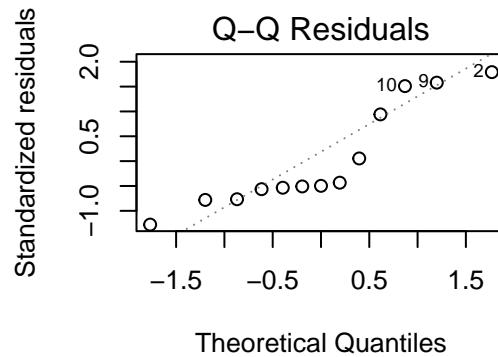
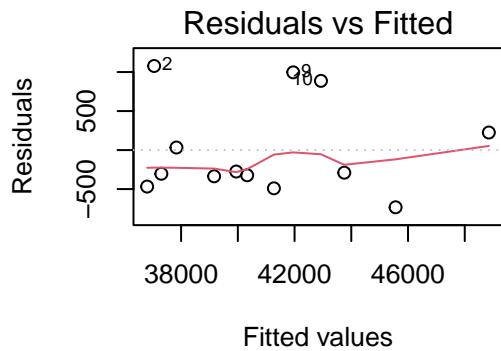
ercot.totalavg.regression <- lm(data = ercot_linear_model_data_totprice,
                                  Mean_Load ~ average_price_tot + percent_renew)

summary(ercot.totalavg.regression)

##
## Call:
## lm(formula = Mean_Load ~ average_price_tot + percent_renew, data = ercot_linear_model_data_totprice)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -733.5 -336.3 -289.7  225.1 1077.8
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 11580.9    4729.9   2.448  0.03435 *
## average_price_tot 2300.1     511.5   4.497  0.00115 **
## percent_renew     353.6      53.7   6.585 6.2e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 667.8 on 10 degrees of freedom
## Multiple R-squared:  0.9723, Adjusted R-squared:  0.9668
## F-statistic: 175.5 on 2 and 10 DF,  p-value: 1.632e-08

```

```
par(mfrow = c(2,2), mar=c(4,4,4,4))
plot(ercot.totalavg.regression)
```



```
par(mfrow = c(1,1))
```

Summary and Conclusions

References

<add references here if relevant, otherwise delete this section>