



What factors influence the price of electricity?

GT Data Analytics - Project 1
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Overview

Goal:

Compare the relationships between electricity price and several factors: temperature, states, market sectors, time, and consumption.

Another factor that emerged during research was regional differences within the US.

Process:

Get data

Clean/merge data

Create numerical summaries

Create visualizations

Write analysis

Correlations:

Electricity Price + Temperature (John)

Electricity Price + State Sectors (Frances)

Electricity Price + Time (Carol)

Temperature + Consumption* (Hugh)

**We used retail sales of electricity by utilities as a proxy.*

APIs / Datasets:

US Energy Information Administration (EIA)

<https://www.eia.gov/opendata/>

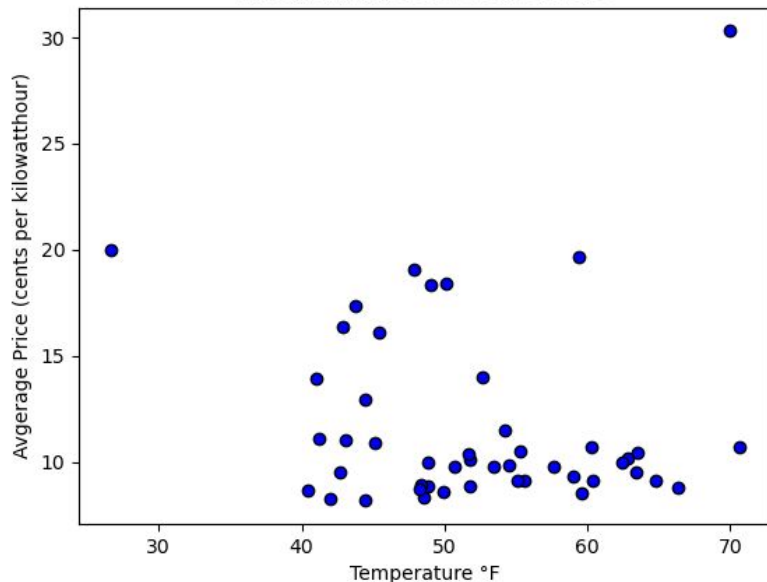
Current Results (weather and science facts)

<https://www.currentresults.com/Weather/US/average-annual-state-temperatures.php>

Electricity Price + Temperature



Electricity Price by Temperature

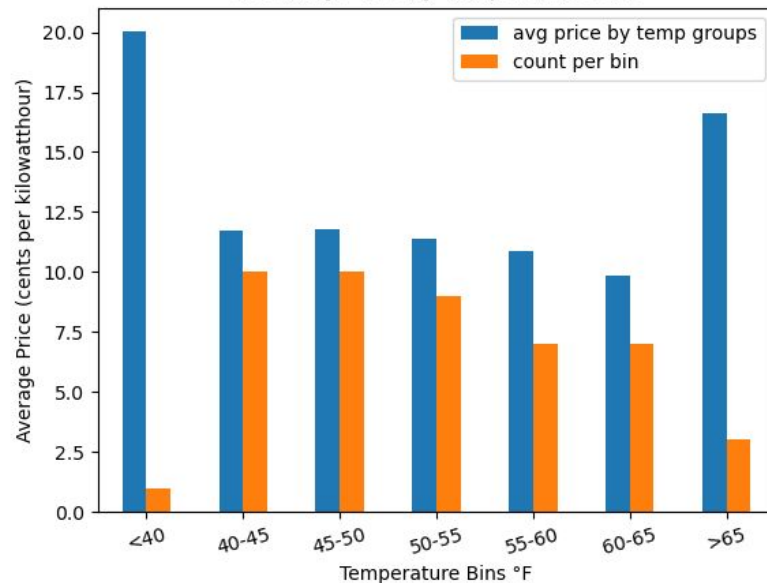


A standard scatter plot shows little to no correlation between electricity price and temperature for 50 states.

- Hawaii and Alaska are notable outliers, intuitively.

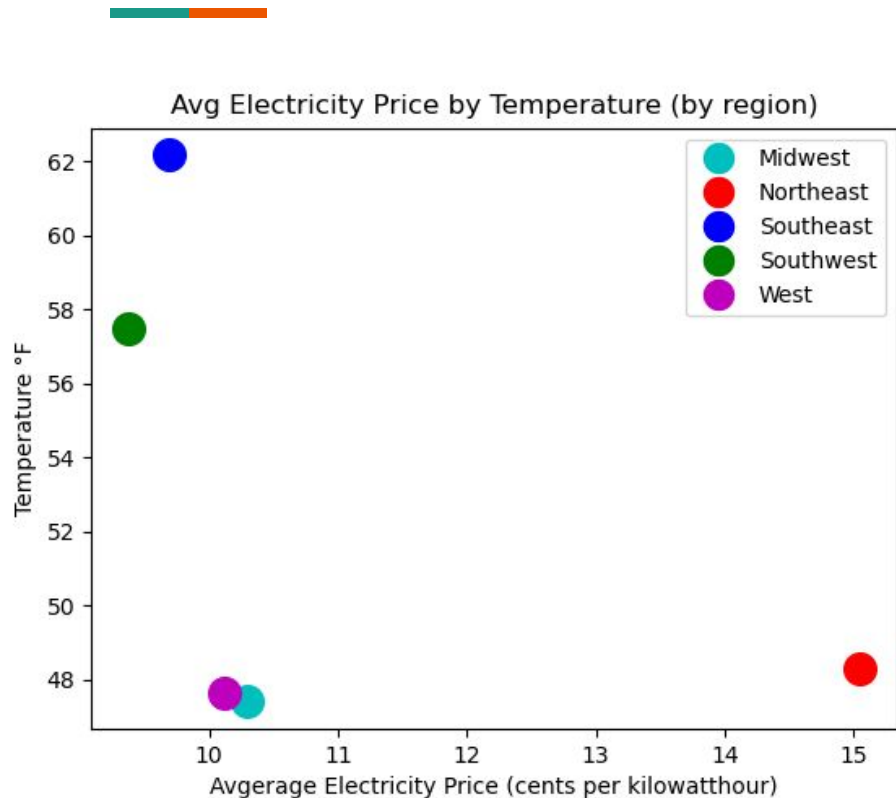
Temperature is binned by five degree increments for simpler visual comparison.

Electricity Price by Temperature Bins





Electricity Price + Temperature



When states were categorized by region, the average retail price for states in the northeast was significantly higher than the other fairly consistently priced regions.

Electricity Price + Temperature

```
1 final_data.head()
```

| | stateID | stateDescription | region | population | Avg °F | Avg °C | Rank | period | average-retail-price | average-retail-price-units | temperature bins |
|---|---------|------------------|-----------|------------|--------|--------|------|--------|----------------------|----------------------------|------------------|
| 0 | AL | Alabama | Southeast | 5,049,846 | 62.8 | 17.1 | 7.0 | 2021.0 | 10.18 | cents per kilowatthour | 60-65 |
| 2 | AZ | Arizona | Southwest | 7,264,877 | 60.3 | 15.7 | 10.0 | 2021.0 | 10.73 | cents per kilowatthour | 60-65 |
| 3 | AR | Arkansas | Southeast | 3,028,122 | 60.4 | 15.8 | 9.0 | 2021.0 | 9.10 | cents per kilowatthour | 60-65 |
| 4 | CA | California | West | 39,142,991 | 59.4 | 15.2 | 12.0 | 2021.0 | 19.65 | cents per kilowatthour | 55-60 |
| 5 | CO | Colorado | West | 5,811,297 | 45.1 | 7.3 | 39.0 | 2021.0 | 10.90 | cents per kilowatthour | 45-50 |

Above shows the DataFrame used in this portion of the project.

The right code snippet shows importing, merging, and cleaning regional data as well as the groupby numerical analysis for the summary dataframe.

```
1 # Import region data & merge with existing data
2 region = "data/region_data.csv"
3 region_df = pd.read_csv(region)
4 complete_data = pd.merge(region_df, all_data2, how="inner", on=["stateDescription"])
5
6 # Clean merged data & drop outliers (Alaska & Hawaii)
7 renamed_data = complete_data.rename(columns={"stateID_x": "stateID"})
8 drop_dup_column = renamed_data.drop(columns="stateID_y")
9 final_data = drop_dup_column.drop([1, 10])
10
11 # Group data by region and calculate average Avg °F & average-retail-price
12 region_temp = clean_df.groupby(["region"]).mean()["Avg °F"]
13 region_price = clean_df.groupby(["region"]).mean()["average-retail-price"]
14
15 # Create dictionary & dataframe respectively
16 region_dict = {
17     "average temperature °F": region_temp,
18     "average retail price (cents per kilowatthour)": region_price,
19 }
20
21 regions_df = pd.DataFrame(region_dict)
22 regions_df
```

| | average temperature °F | average retail price (cents per kilowatthour) |
|-----------|------------------------|---|
| region | | |
| Midwest | 47.400000 | 10.291111 |
| Northeast | 48.281818 | 15.050909 |
| Southeast | 62.180000 | 9.885000 |
| Southwest | 57.500000 | 9.365000 |
| West | 47.644444 | 10.121111 |

Electricity Price + State Sectors

```
base_url = "https://api.eia.gov/v2/electricity/retail-sales/data"
```

```
X_Params = {
    "api_key": api_key,
    "frequency": "monthly",
    "data[0]": "price",
    "facets": {},
    "start": "2021-01",
    "end": "2021-12",
    "sort[0][column]": "period",
    "sort[0][direction]": "desc",
    "offset": 0,
    "length": 5000
}
```

```
response = requests.get(base_url, params=X_Params).json()
```

```
sectorDF = pd.DataFrame(response["response"]["data"])
display(sectorDF.head())
print(f"Number of Rows: {len(sectorDF)}")
```

| | period | stateid | stateDescription | sectorid | sectorName | price | price-units |
|---|---------|---------|------------------|----------|----------------|-------|------------------------|
| 0 | 2021-12 | IN | Indiana | OTH | other | NaN | cents per kilowatthour |
| 1 | 2021-12 | US | U.S. Total | TRA | transportation | 10.49 | cents per kilowatthour |
| 2 | 2021-12 | US | U.S. Total | RES | residential | 13.72 | cents per kilowatthour |
| 3 | 2021-12 | US | U.S. Total | OTH | other | NaN | cents per kilowatthour |
| 4 | 2021-12 | US | U.S. Total | IND | industrial | 7.06 | cents per kilowatthour |

Number of Rows: 4464

```
sectorDF.dropna(subset=["price"], inplace=True, how="all")
```

```
len(sectorDF)
```

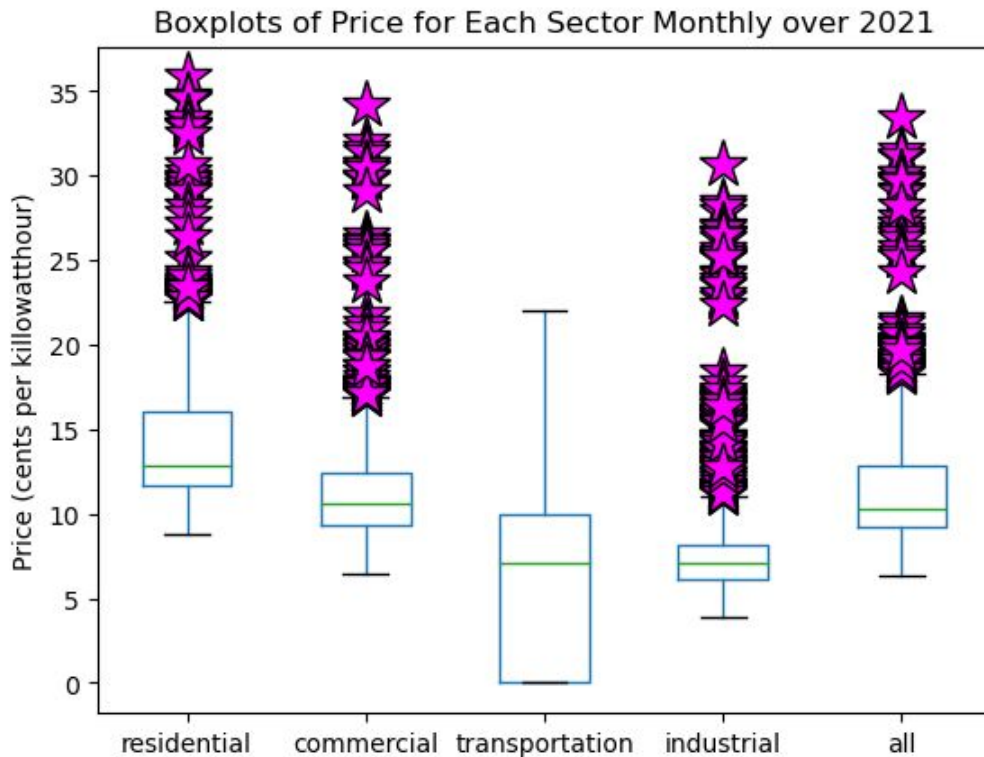
3720

I solely used the retail sales data available through the EIA for my analysis. I honed in on 2021 because that was the most recent year with data for every month at the beginning of our project.

Electricity Price + State Sectors



In 2021, the price of electricity across all states varied somewhat by sector. The spread for the residential sector is higher than other sectors, while the transportation sector is lower than the others.

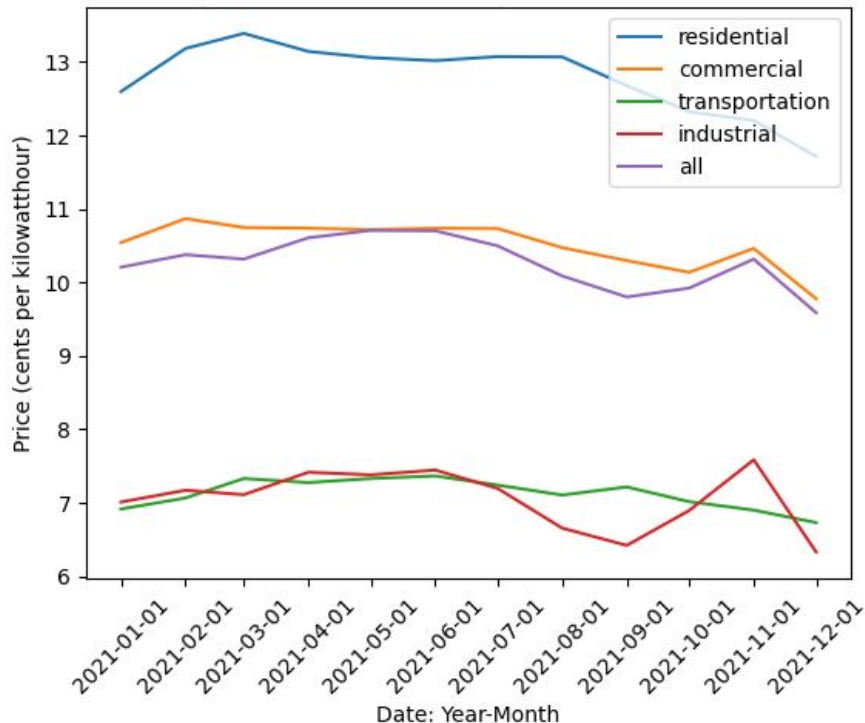


Electricity Price + State Sectors

The rate increases and decreases in the median price of electricity across sectors mirrored each other fairly well in 2021.

Like we saw in the box plots, the residential sector has higher prices than other sectors, and transportation/industrial sectors have lower prices.

Median Electricity Price over all States by Market Sector Over Time in 2021



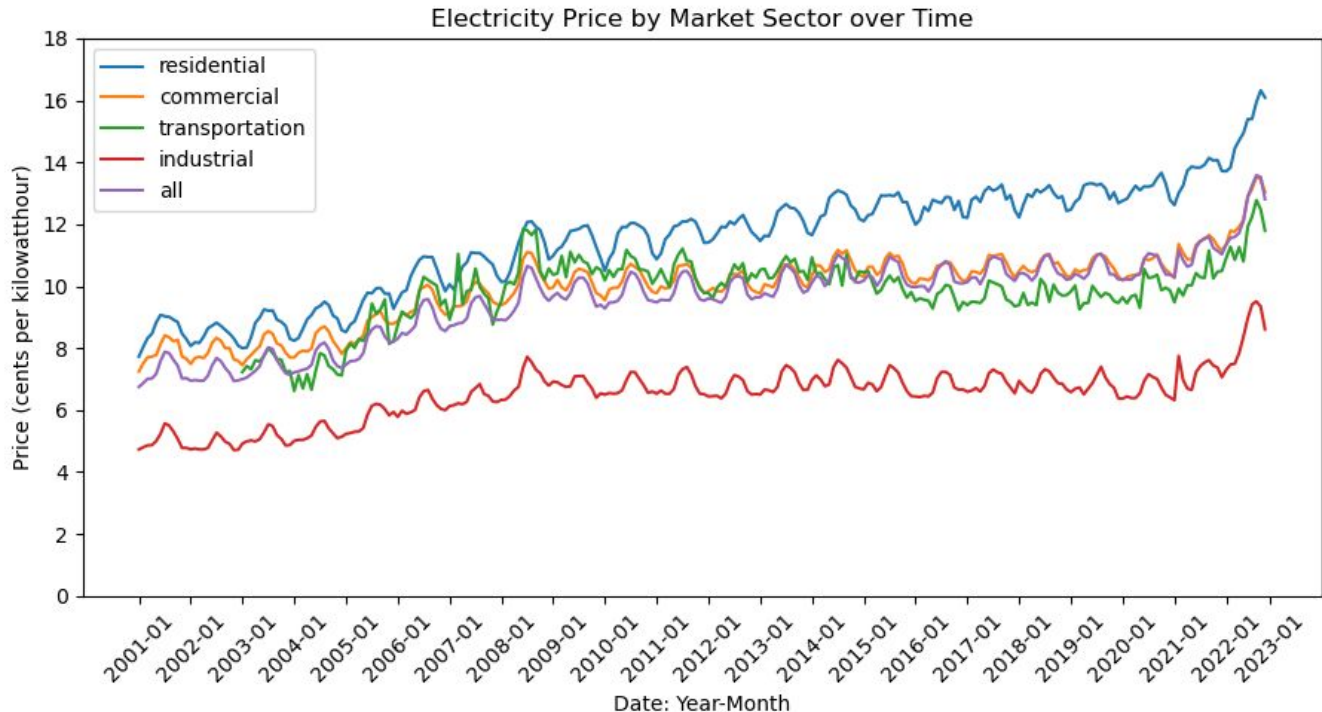


Electricity Price + Time

Industrial customers tend to pay less for electricity than other sectors.

Residential customers tend to pay more for electricity than other sectors.

Residential customers also saw a greater increase in price over time.





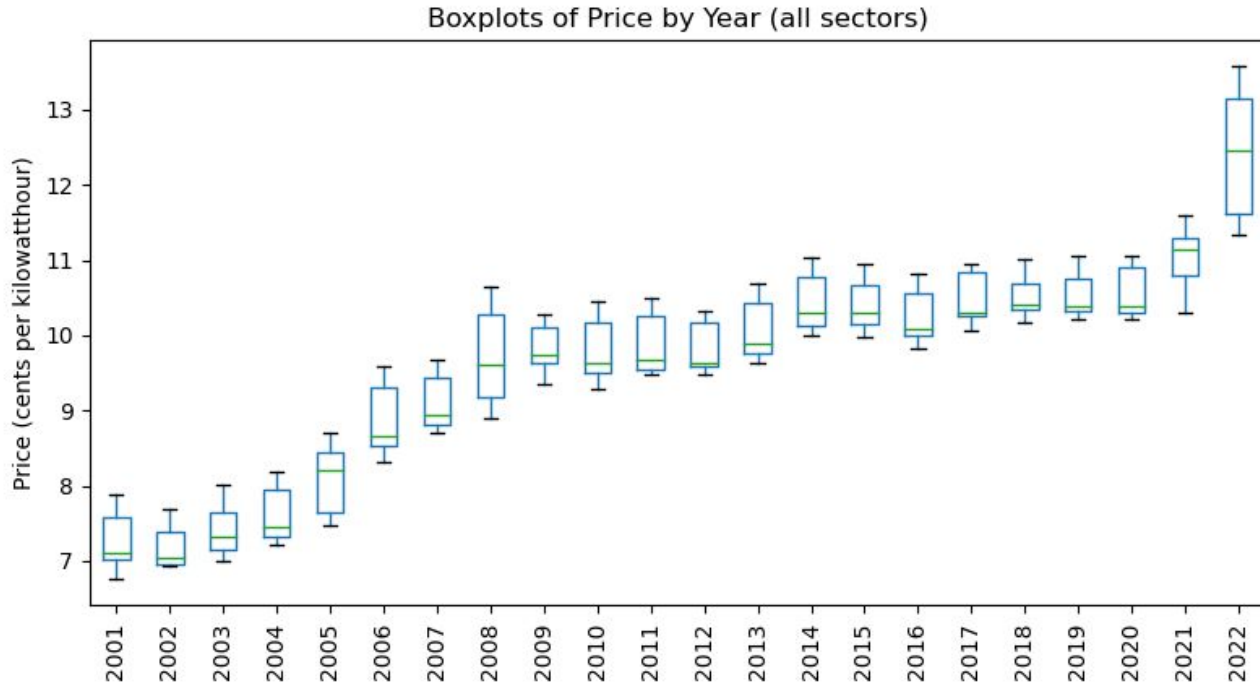
Electricity Price + Time

The boxplots show the median and spread of the prices for each year.

In general, the median price (green line) increases over time.

Most years have similar variability in price, with notably larger spreads in 2008 and 2022.

Also to note: the large increase in median price from 2020 - 2022.



Electricity Price + Time

```
In [15]: year_summary_stats = time_cost_all.groupby(["year"]).agg(["mean", "median", "var", "std", "sem"])(["price"])
         year_summary_stats
```

```
Out[15]:
```

| | mean | median | var | std | sem |
|------|-----------|--------|----------|----------|----------|
| year | | | | | |
| 2001 | 7.268333 | 7.100 | 0.148342 | 0.385152 | 0.111184 |
| 2002 | 7.175000 | 7.040 | 0.075373 | 0.274541 | 0.079253 |
| 2003 | 7.416667 | 7.330 | 0.123970 | 0.352093 | 0.101641 |
| 2004 | 7.595000 | 7.440 | 0.125100 | 0.353695 | 0.102103 |
| 2005 | 8.107500 | 8.210 | 0.211893 | 0.460319 | 0.132883 |
| 2006 | 8.860833 | 8.670 | 0.207081 | 0.455062 | 0.131365 |
| 2007 | 9.104167 | 8.950 | 0.126863 | 0.356178 | 0.102820 |
| 2008 | 9.705000 | 9.600 | 0.401100 | 0.633325 | 0.182825 |
| 2009 | 9.804167 | 9.735 | 0.101808 | 0.319074 | 0.092109 |
| 2010 | 9.797500 | 9.635 | 0.161711 | 0.402134 | 0.116086 |
| 2011 | 9.863333 | 9.680 | 0.158297 | 0.397866 | 0.114854 |
| 2012 | 9.815833 | 9.640 | 0.110263 | 0.332059 | 0.095857 |
| 2013 | 10.043333 | 9.890 | 0.148188 | 0.384952 | 0.111126 |
| 2014 | 10.420833 | 10.305 | 0.127990 | 0.357757 | 0.103276 |
| 2015 | 10.385000 | 10.300 | 0.115391 | 0.339692 | 0.098061 |
| 2016 | 10.242500 | 10.095 | 0.113384 | 0.336726 | 0.097204 |
| 2017 | 10.458333 | 10.310 | 0.109070 | 0.330257 | 0.095337 |
| 2018 | 10.506667 | 10.410 | 0.078861 | 0.280821 | 0.081066 |
| 2019 | 10.516667 | 10.385 | 0.090842 | 0.301401 | 0.087007 |
| 2020 | 10.560833 | 10.385 | 0.114227 | 0.337974 | 0.097565 |
| 2021 | 11.071667 | 11.150 | 0.156397 | 0.395471 | 0.114163 |
| 2022 | 12.436000 | 12.460 | 0.754071 | 0.868373 | 0.274604 |

We used the API key and parameters from the US Energy Information Administration (EIA) API to get electricity price data, then removed rows that were missing data for the price of electricity.

We used .agg to create a dataframe of summary statistics for the price of electricity by year (shown) and by month (not shown).

<https://pandas.pydata.org/docs/reference/api/pandas.DataFrame.agg.html>

```
In [3]: # Set the base url
elec_url = "https://api.eia.gov/v2/electricity/retail-sales/data/"

# Define the search parameters
parameters = {"api_key": elec_key,
              "frequency": "monthly",
              "data[0]": "price",
              "facets[sectorid][]": ["ALL", "COM", "IND", "RES", "TRA", "OTH"],
              "facets[stateid][]": "US",
              "start": "2001-01",
              "end": "2022-10",
              "sort[0][column]": "period",
              "sort[0][direction]": "asc",
              "length": "5000",}

# Get the json data from the API
time_cost = requests.get(elec_url, params = parameters).json()

# Create a DataFrame from the json data
time_cost = pd.DataFrame(time_cost["response"]["data"])
time_cost.head()
```

```
Out[3]:
```

| | period | stateid | stateDescription | sectorid | sectorName | price | price-units |
|---|---------|---------|------------------|----------|----------------|-------|------------------------|
| 0 | 2001-01 | US | U.S. Total | ALL | all sectors | 6.75 | cents per kilowatthour |
| 1 | 2001-01 | US | U.S. Total | TRA | transportation | NaN | cents per kilowatthour |
| 2 | 2001-01 | US | U.S. Total | RES | residential | 7.73 | cents per kilowatthour |
| 3 | 2001-01 | US | U.S. Total | OTH | other | 6.48 | cents per kilowatthour |
| 4 | 2001-01 | US | U.S. Total | IND | industrial | 4.73 | cents per kilowatthour |

```
In [4]: # Remove rows that are missing prices
time_cost.dropna(subset=["price"], how = "all", inplace = True)
time_cost.head()
```

```
Out[4]:
```

| | period | stateid | stateDescription | sectorid | sectorName | price | price-units |
|---|---------|---------|------------------|----------|-------------|-------|------------------------|
| 0 | 2001-01 | US | U.S. Total | ALL | all sectors | 6.75 | cents per kilowatthour |
| 2 | 2001-01 | US | U.S. Total | RES | residential | 7.73 | cents per kilowatthour |
| 3 | 2001-01 | US | U.S. Total | OTH | other | 6.48 | cents per kilowatthour |
| 4 | 2001-01 | US | U.S. Total | IND | industrial | 4.73 | cents per kilowatthour |
| 5 | 2001-01 | US | U.S. Total | COM | commercial | 7.25 | cents per kilowatthour |

Temperature v. Electricity Consumption per Capita by US State - 2021



United States
overall
total
electricity
consumption
2021

total kWh
consumption

total
population

~3,500,000
million kWh

~332 million
people

~11,500 kWh / capita

source: EIA SEDS ESTCP kWh units

Temperature v. Electricity Consumption per Capita by US State - 2021



Linear regression

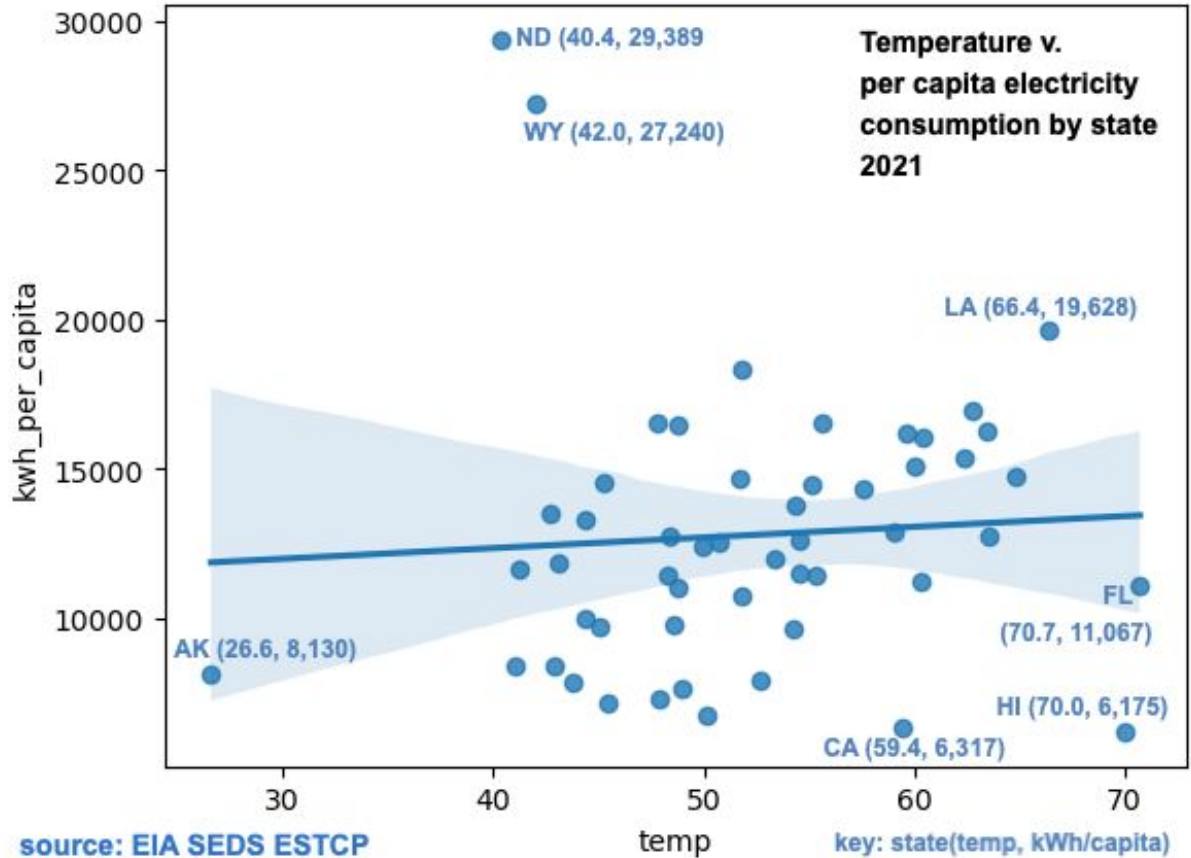
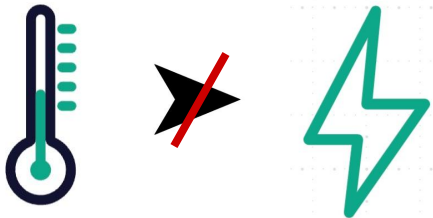
independent variable: temperature

dependent variable: consumption

Coefficient of determination
(R-squared):
0.0045

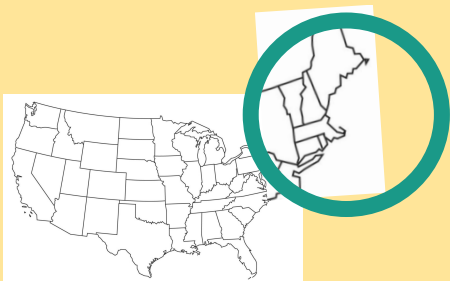
R score (correlation coefficient)
 $\sqrt{R^2} = R = 0.067$

No discernable relationship
between temperature and
electricity consumption by state



Takeaways

John:



Electricity costs more in the Northeast

Frances:

Residential sector price > Transportation/Industrial



Similar rate of change across all sectors.

Carol:



Greatest price increase from 2020 - 2022.

Hugh:

2021

consumption per capita



~11,500 kWh/year



~1,000 kWh/month

Questions, Comments, Concerns?



Temperature v. Electricity Consumption per Capita by US State - 2021

How outliers help to understand limitations of the dataset & weaknesses in the hypothesis that per capita electricity consumption is correlated with temperature

| State | Avg. 2021 temp | kWh per capita | possible reason | corresponding limitation |
|-------|----------------|----------------|---|---|
| CA | 59.4 | 6,317 | high cost(\$0.239/kWh) geographic temp variation higher proliferation of DERs | price varies among states population not evenly distributed according to temperature variation a higher proportion of kWh consumed not accounted for in utility-based retail sales data |

Temperature v. Electricity Consumption per Capita by US State - 2021

| State | Avg. 2021 temp | kWh per capita | possible reason | corresponding limitation |
|-------|----------------|----------------|--|--|
| HI | 70.0 | 6,175 | high cost(\$0.30/kWh) less temp variation higher wind speeds | price varies among states less work for HVAC less work for HVAC |
| FL | 70.7 | 11,067 | lowish cost(\$0.125/kWh) more seasonal variation of temp high humidity | price varies among states more work in summer & winter for HVAC all else equal, more work for HVAC |
| LA | 66.4 | 19,628 | low cost(\$0.117/kWh) more seasonal variation of temp high humidity | price varies among states more work in summer & winter for HVAC all else equal, more work for HVAC |

Temperature v. Electricity Consumption per Capita by US State - 2021

| State | Avg. 2021 temp | kWh per capita | possible reason | corresponding limitation |
|-------|----------------|----------------|---|---|
| WY | 40.4 | 27,240 | very low cost(\$0.109) more seasonal temp variation | price varies among states more work in summer & winter for HVAC |
| ND | 42.0 | 29,398 | very low cost(\$0.109) more seasonal temp variation | price varies among states more work in summer & winter for HVAC |
| AK | 26.6 | 8,130 | high cost(\$.226) less electricity used in building energy | price varies among state more thermal energy sources for building energy |