



# Market the Market

The Nightwatch

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# Demo

```
In [*]: state_code = input("Please Enter the State!")  
while state_code not in {'TX', 'CO', 'GA'}:  
    state_code = input("Please Enter a Valid State Code!")
```

Please Enter the State!TA

Please Enter a Valid State Code:

```
In [ ]: get_daily_traffic_avg(model, state_code)
```

```
In [ ]:
```



# Overview

To find the most ideal place to start a new supermarket, we first use data visualization tools including carto, to visualize the patterns of the dataset, and use different statistical model to predict the traffic, as an indicator of profitability.





## Question & Hypothesis

Can different factors such as store location, community population, competitor count, customer income... predict customer traffic?



# Outsourced Datasets

- Location dataset
  - fipsCode → longitude and latitude
    - <https://www.quora.com/Where-are-latitude-longitude-coordinates-for-all-census-block-FIPS-codes-available>
- Income dataset
  - fipsCode → zipcode
  - Zip code → income
  - <https://www.kaggle.com/goldenoakresearch/us-household-income-stats-geo-locations>



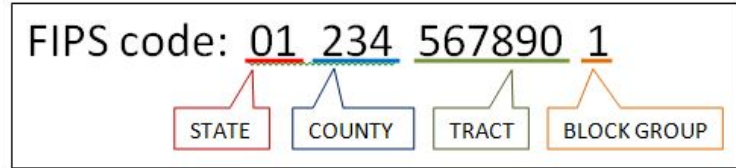
# Outsourced Datasets

- Location dataset

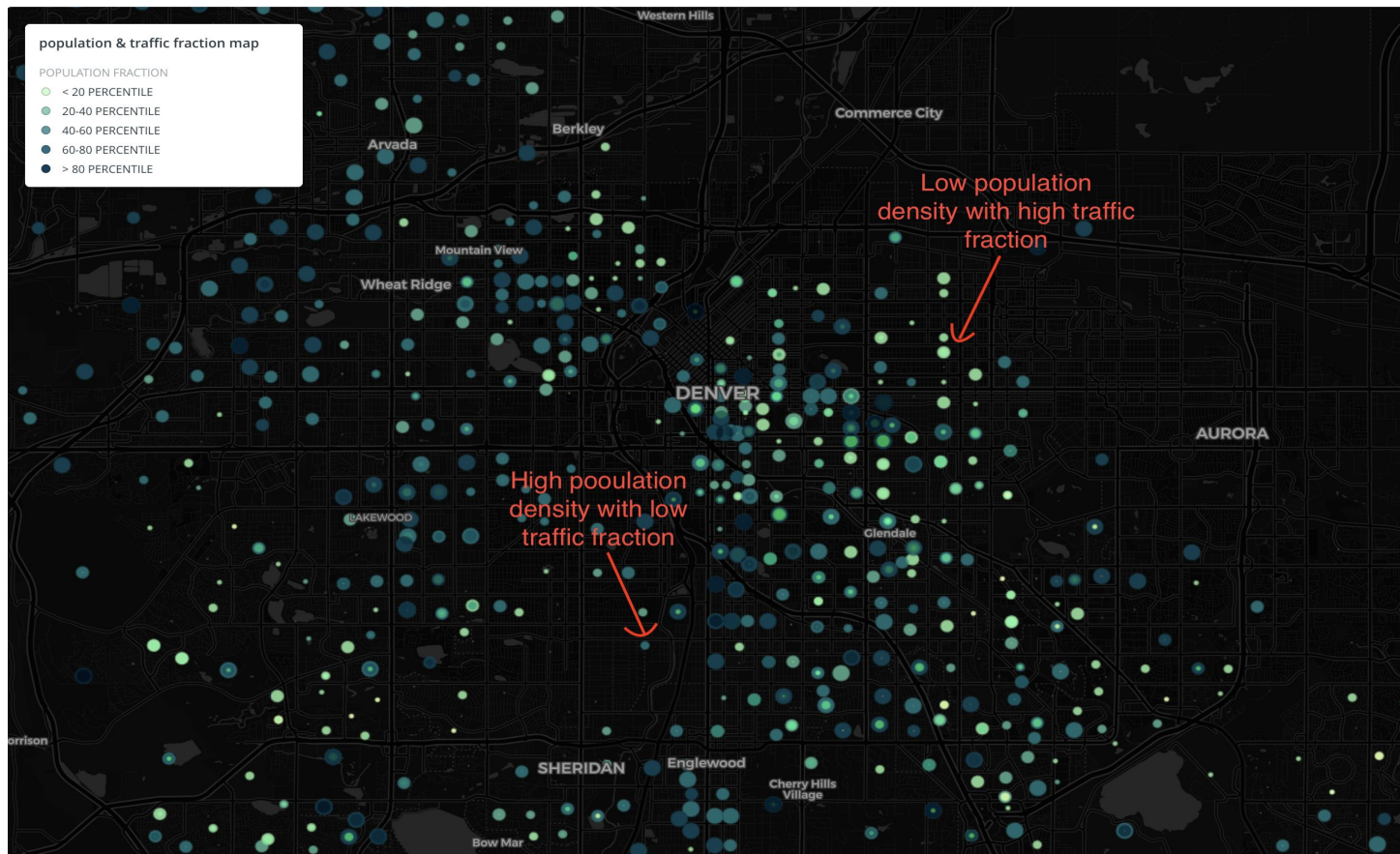
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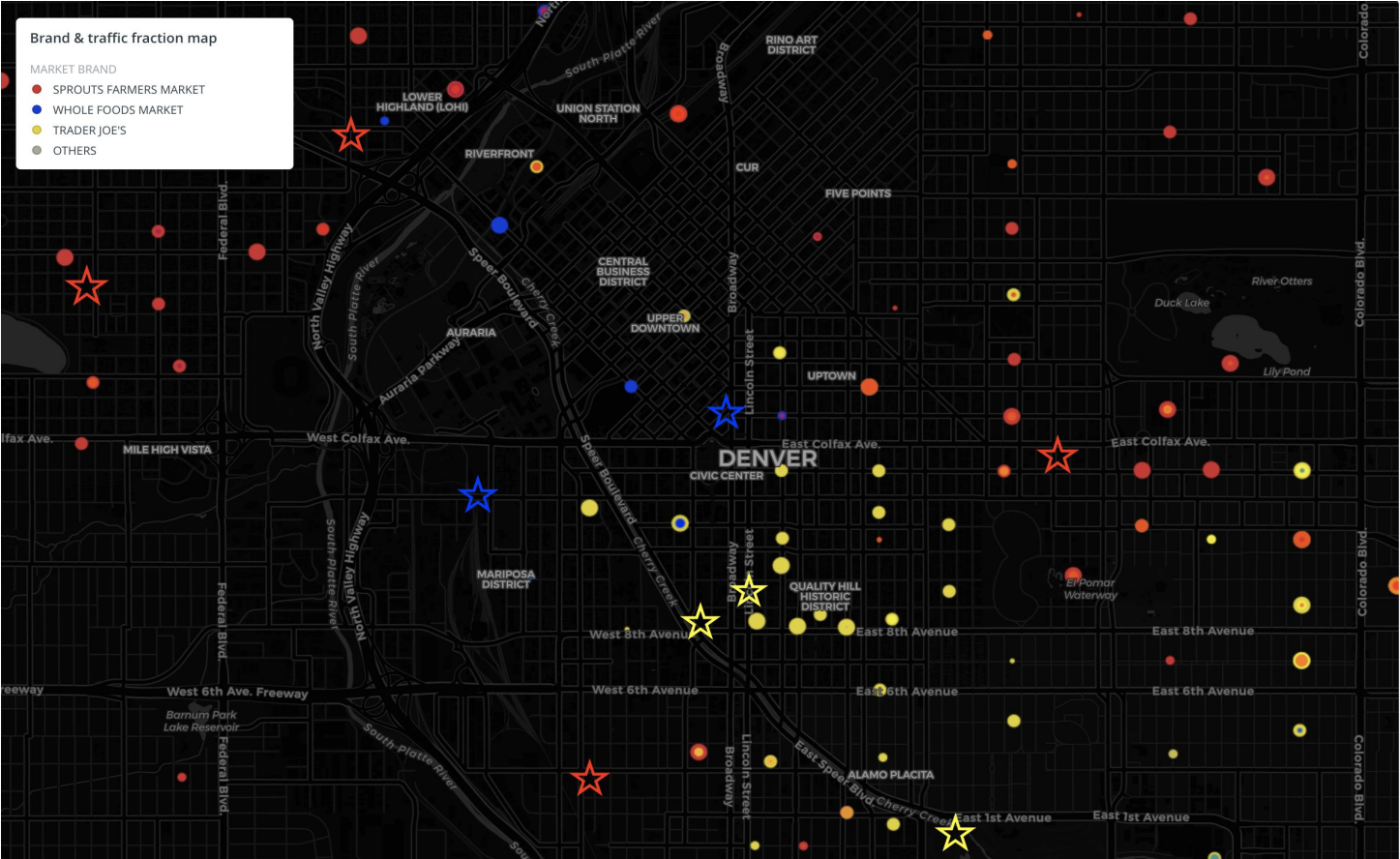


## Population density & Traffic fraction





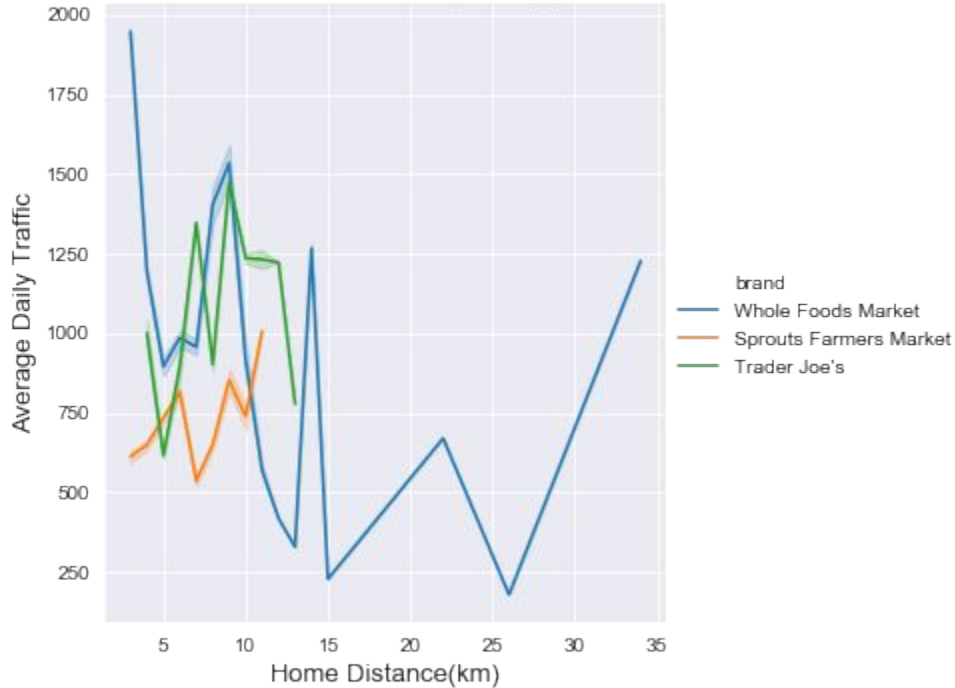
# Store location & Traffic fraction



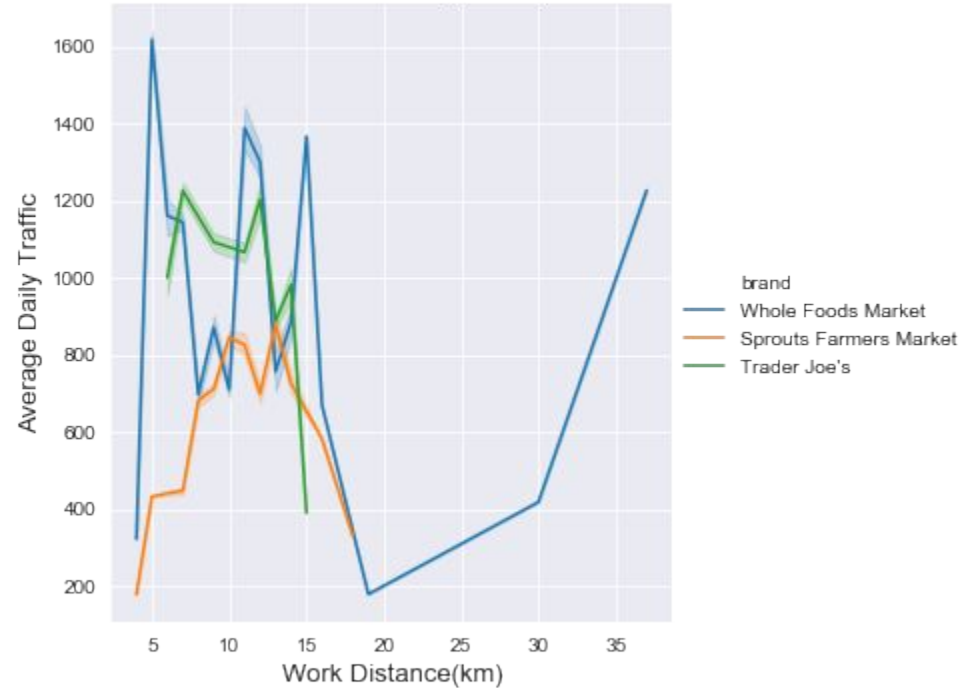
## Average Daily Traffic vs Distance



### Home Distance vs Avg Daily Traffic



### Work Distance vs Avg Daily Traffic



# The Number of Each Brand by State



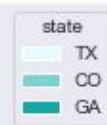
Number of Market

Brand

Whole Foods Market

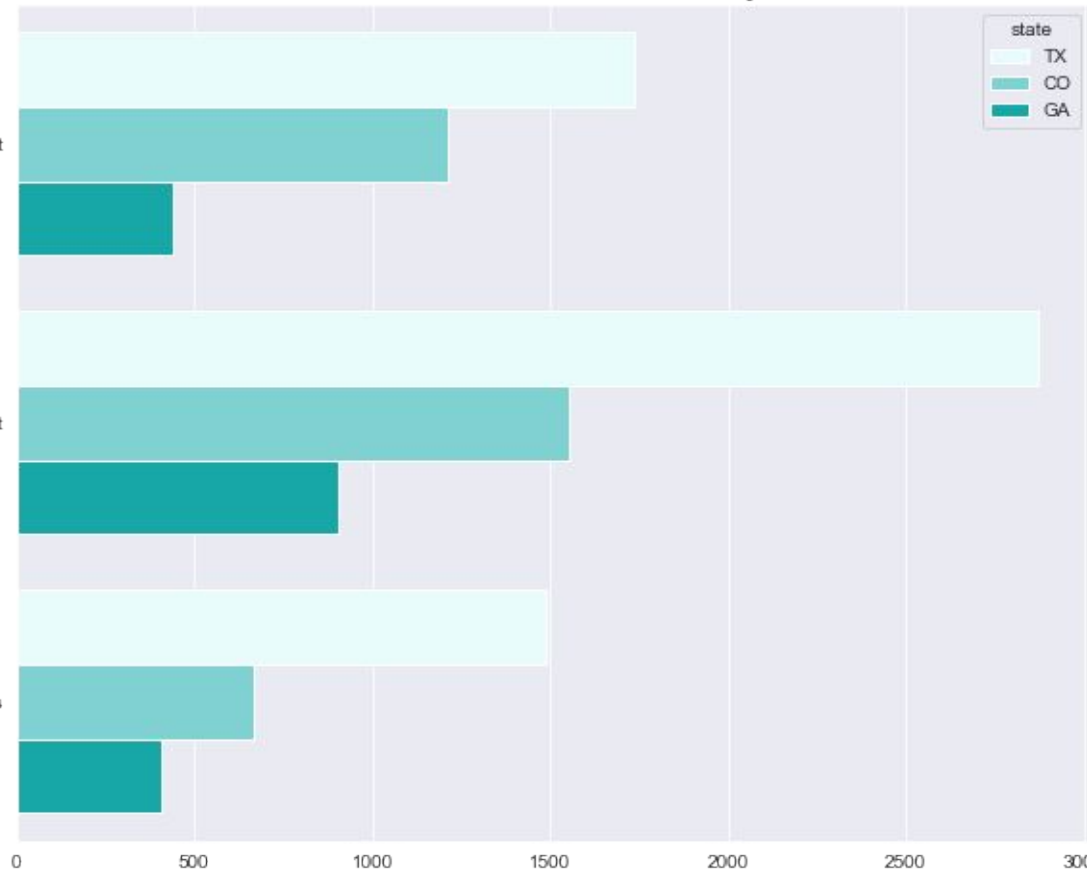
Sprouts Farmers Market

Trader Joe's

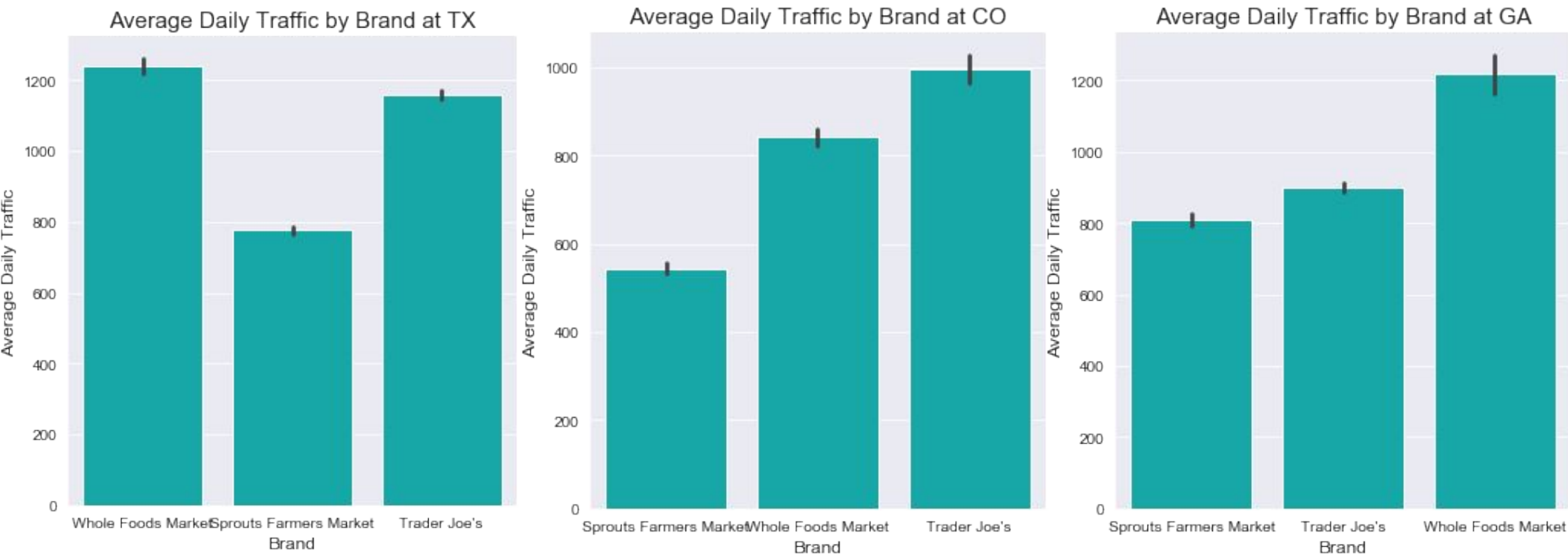


0 500 1000 1500 2000 2500 3000

Count



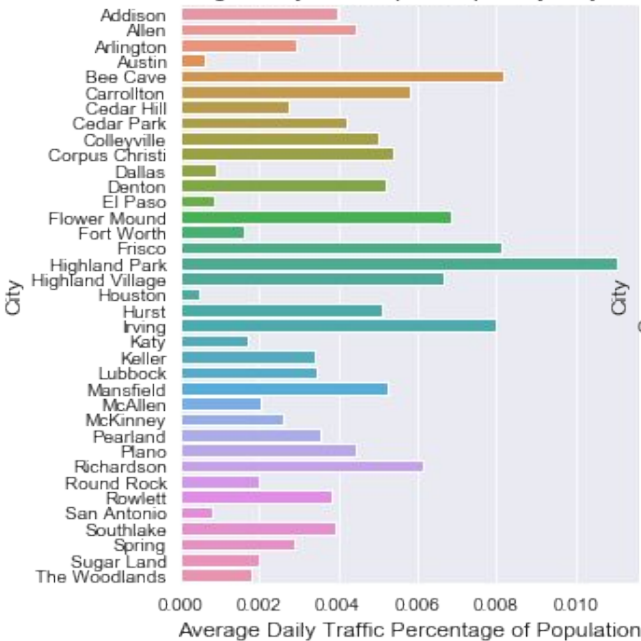
## Average Daily Traffic by Brand



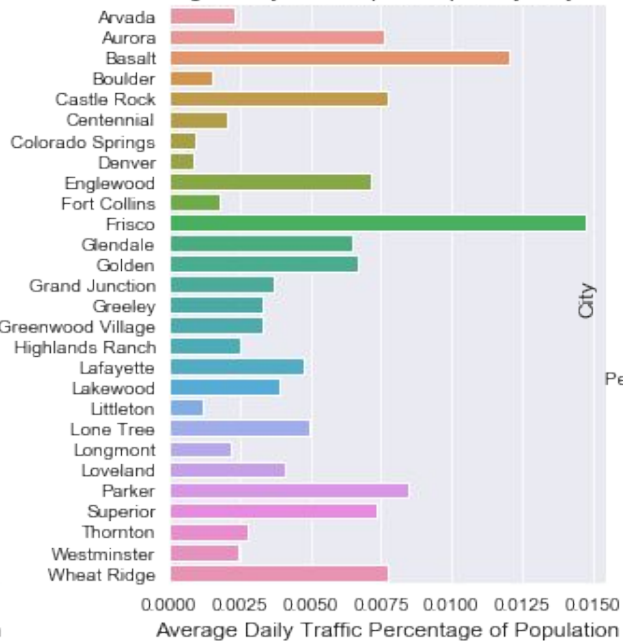
# Average Daily Traffic in Percentage of Population



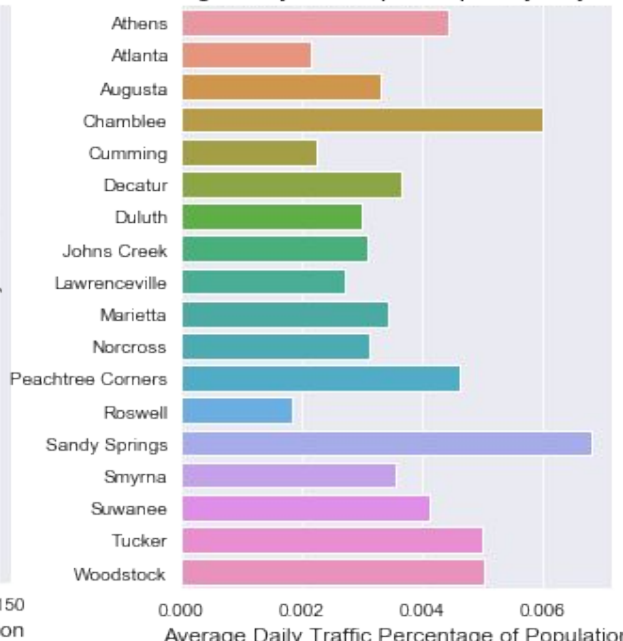
Average Daily Traffic per Capita by City at TX



Average Daily Traffic per Capita by City at CO



Average Daily Traffic per Capita by City at GA



# Machine Learning Model



# Model Setting: Regressor!

- Input (11)
  - State (3)
  - Home-market distance median
  - Home-market distance 25th percentile
  - Home-market distance 75th percentile
  - Workplace-market distance median
  - Workplace-market distance 25th percentile
  - Workplace-market distance 75th percentile
  - Community household annual income median
  - Community household annual income mean
- Output
  - Average traffic per day



# Data Preprocessing

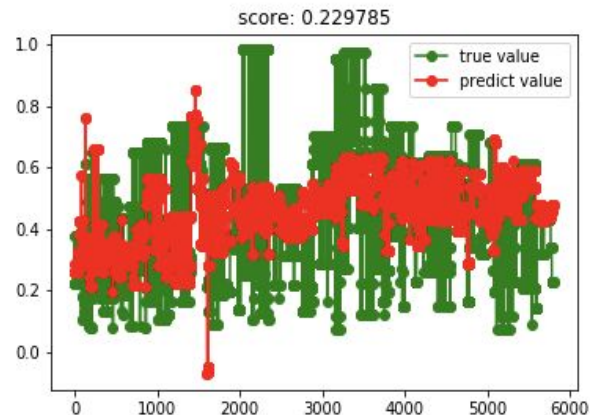
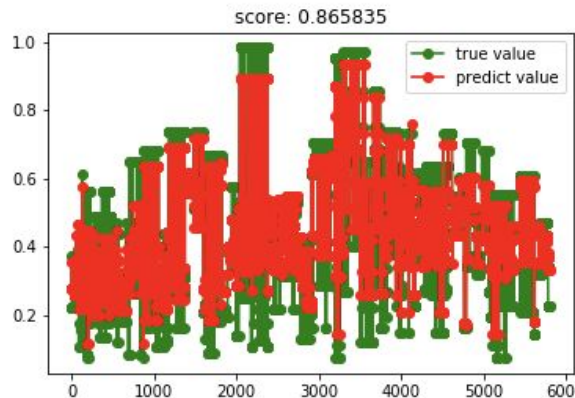
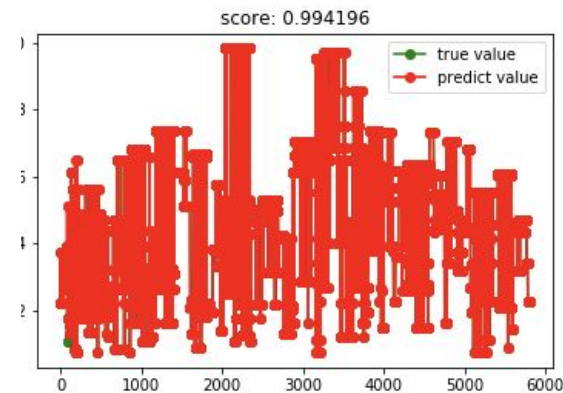
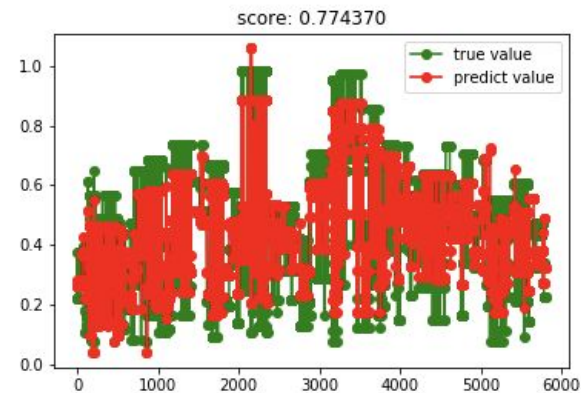
daily_traffic_avg	/2000.
state	one-hot
Distance (any)	/10.
Market count by city	/10.
Income (any)	/1000.





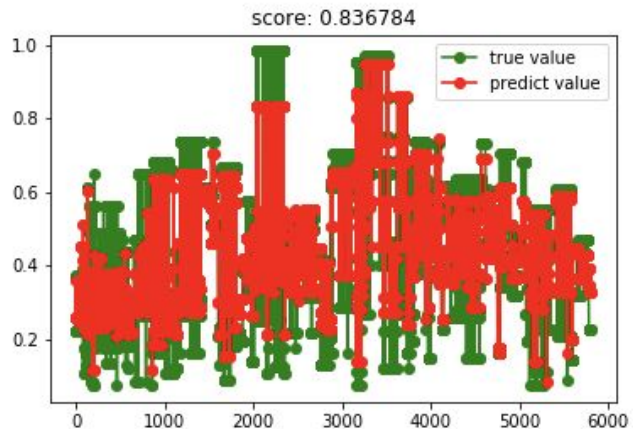
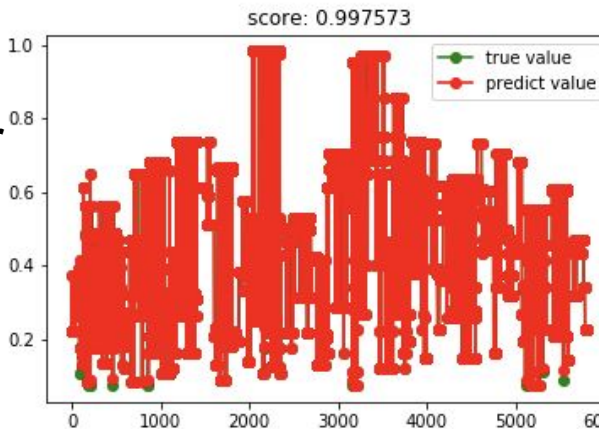
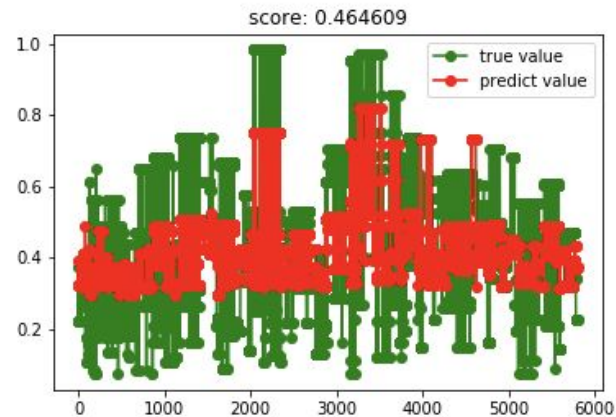
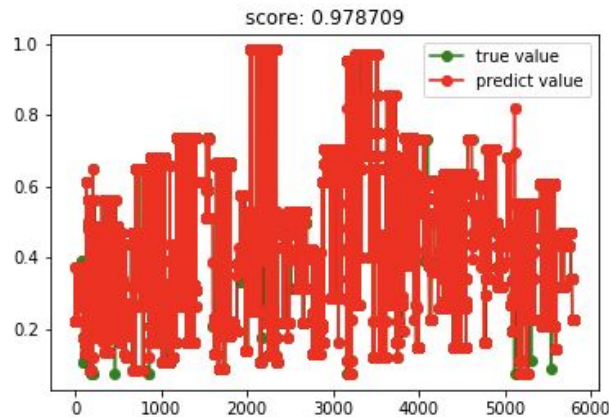
# Different Model?

- **SVR** 0.774370
- **Decision Tree Regressor** 0.994196
- **Gradient boosting regressor** 0.865835
- **Linear Regression** 0.229785



# Different Model?

- **KNeighborsRegressor**  
0.978709
- **AdaBoostRegressor**  
0.464609
- **RandomForestRegressor**  
0.997573
- **GradientBoostingRegressor**  
0.836784





## What we learn by models?

- Our problem is more nonlinear than linear:

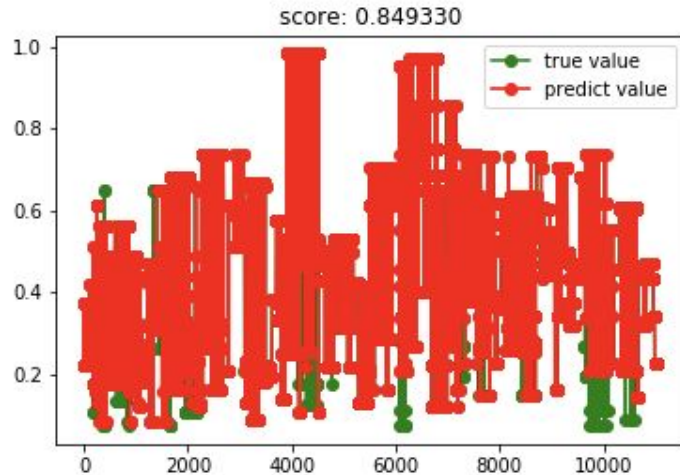
Linear Regression & Adaboost are miserable.

- Decision Tree family is good at this:

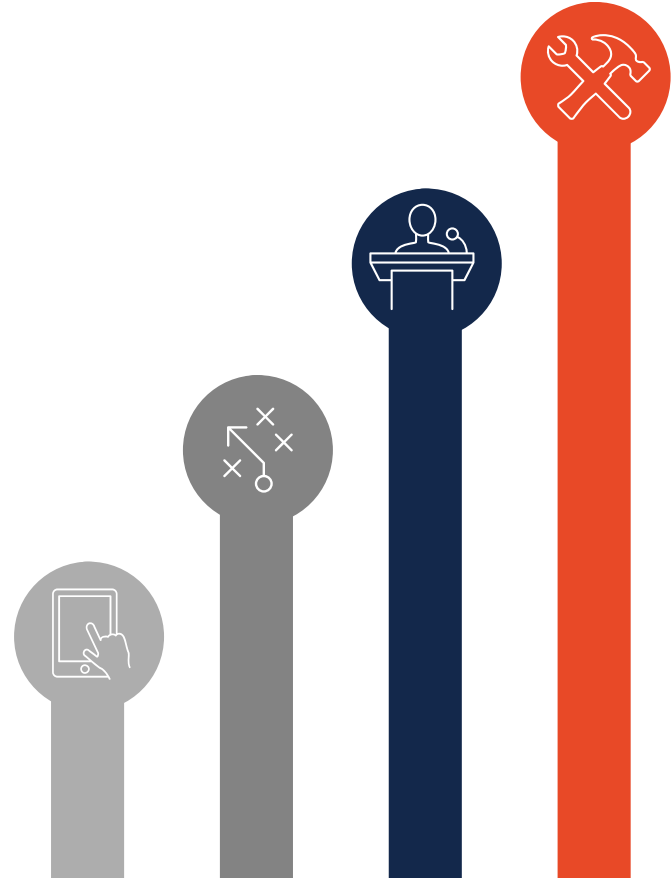
They learn to form a knowledge map of states and cities, and easily leverage it by guessing where the incoming data point should fall



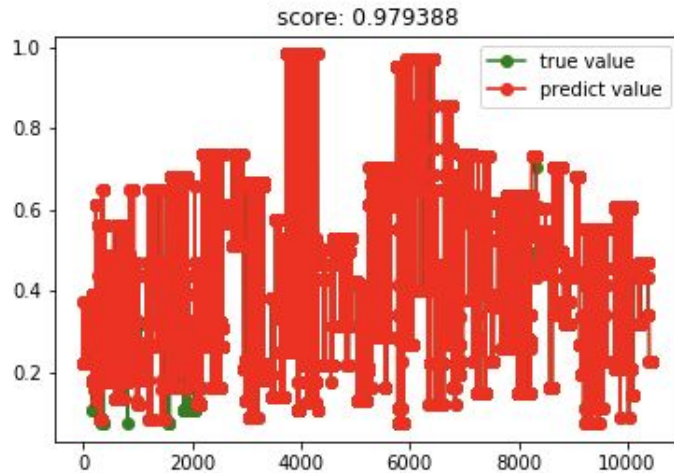
# Data Split Comparison



0.849330  
5%training



# Data Split Comparison



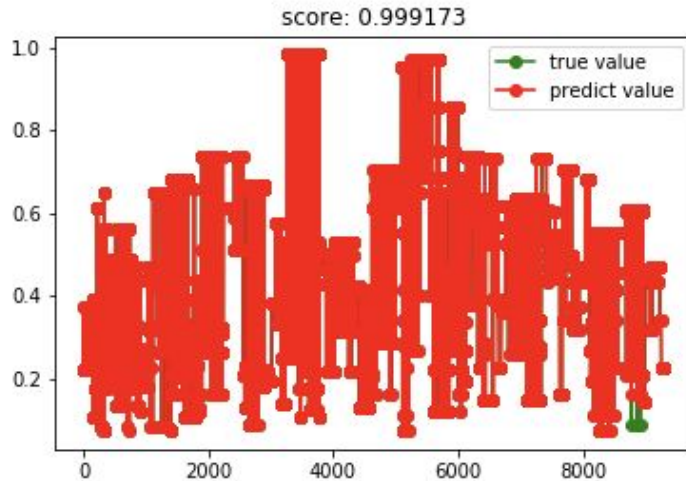
0.979388  
10%training

0.849330  
5%training





# Data Split Comparison



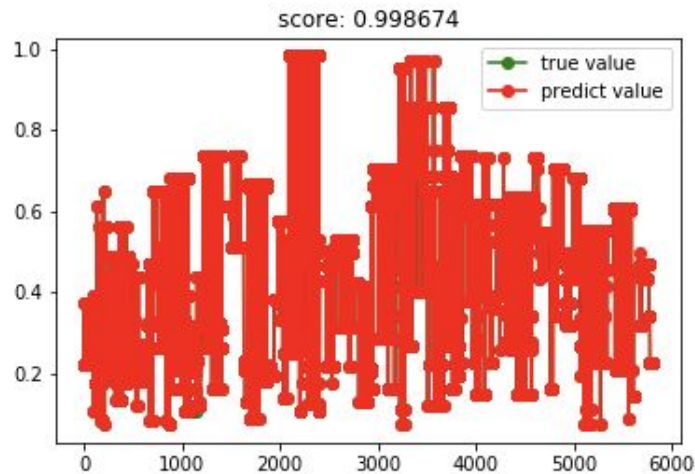
0.849330  
5%training

0.979388  
10%training

0.999173  
20%training



# Data Split Comparison



0.849330  
5%training

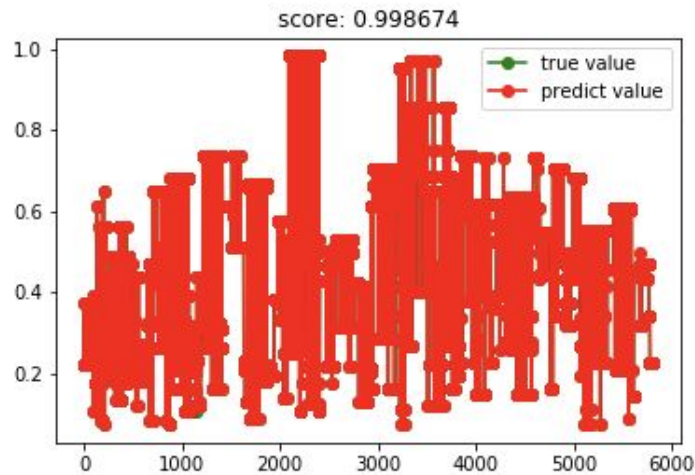
0.979388  
10%training

0.999173  
20%training

0.998674  
50%training



# Random Seed?



0.849330  
5%training

0.979388  
10%training

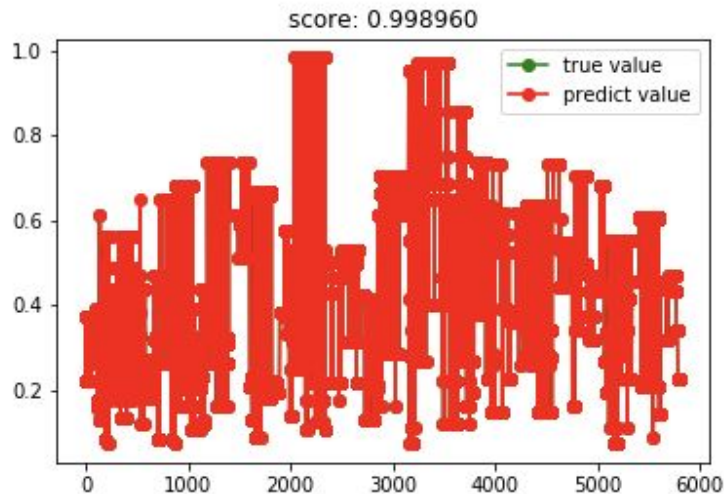
0.999173  
20%training

0.998674  
50%training





# Random Seed?



0.998960  
50%training  
Seed = 100

0.998674  
50%training



0.999173  
20%training



0.979388  
10%training



0.849330  
5%training





## What we know so far?

- Our model is soooo coolll!

Predict well + **Generalize well!!** -> caught some key ideas

- Why they look similar even given different random seed?

Did random sampling, but did not shuffle!

- The model is good, just because it is good! Nothing to do with seed initialization.



## PCA Analysis - Variance

State: 9.56477545e-01 3.72834481e-02 4.93318813e-03 (**high**)

Distance : 4.69759238e-04 2.89794858e-04 2.63631342e-04  
1.32852884e-04 6.97724107e-05 4.10084916e-05

Count of competitors in one city: 2.37527445e-05

Income: 1.18478925e-05 3.39851046e-06 (**low**)



## Go into the future

- Combine Google Map, user simply needs to provide store location to get the predicted customer traffic.
- Expand the database to other industries such as real estate, restaurant.
- Obtain more potential features such as local GDP and local tax to make the model more powerful.



# Thank you.

