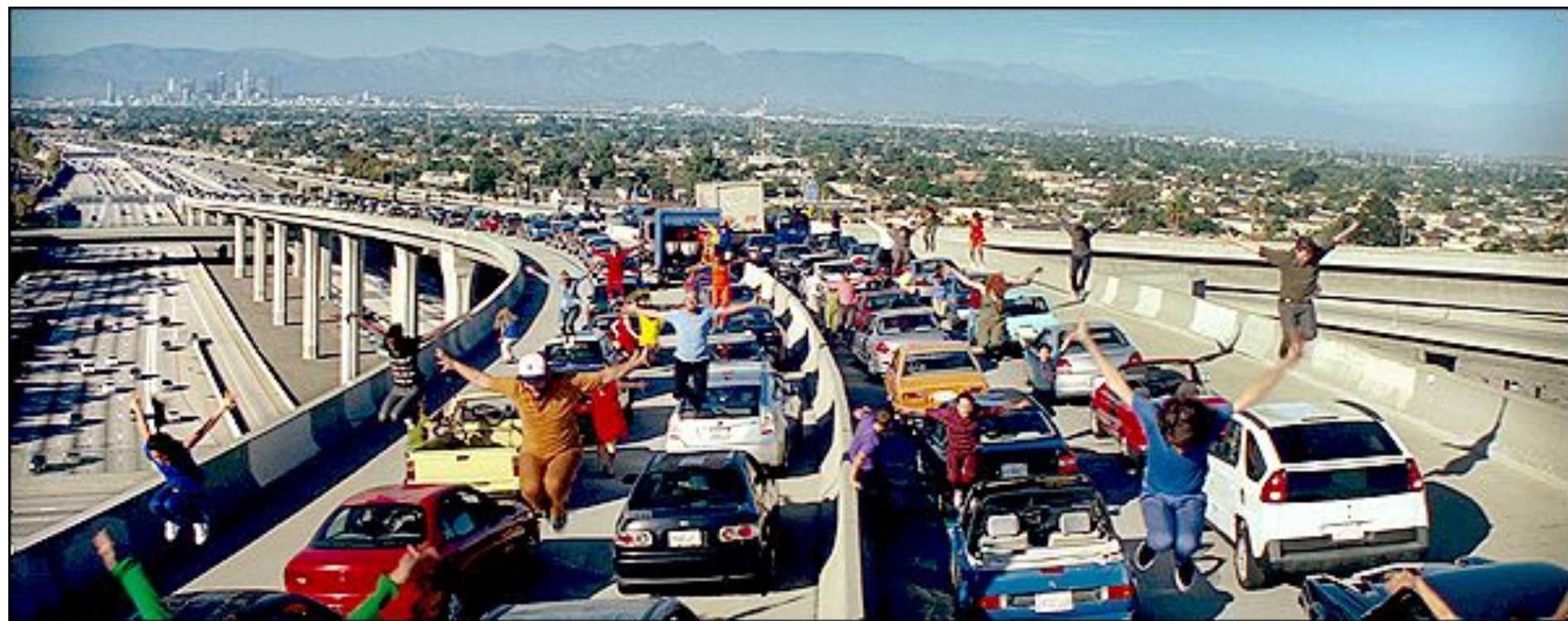


# Analyzing Regional Traffic Count

Team : DC20017

Haoning Deng  
Junfei Hao  
Mit Lalitbhai Patel  
Qinyu Wei



**But in reality...the horrible series...**

# **The Rush Hour**



# Truth about DC traffic



## CONGESTION LEVEL NOW

● **47%**

↓ **9%** less than average at this time

## TRAFFIC JAMS NOW

**423**

total count

**475.8 km**

total length

**WORLD RANK  
2019**

**141**

## TIME LOST IN RUSH HOUR - PER TRIP

How much extra time was spent driving in rush hour?



**+14 min**

per 30 min trip  
in the morning

**+18 min**

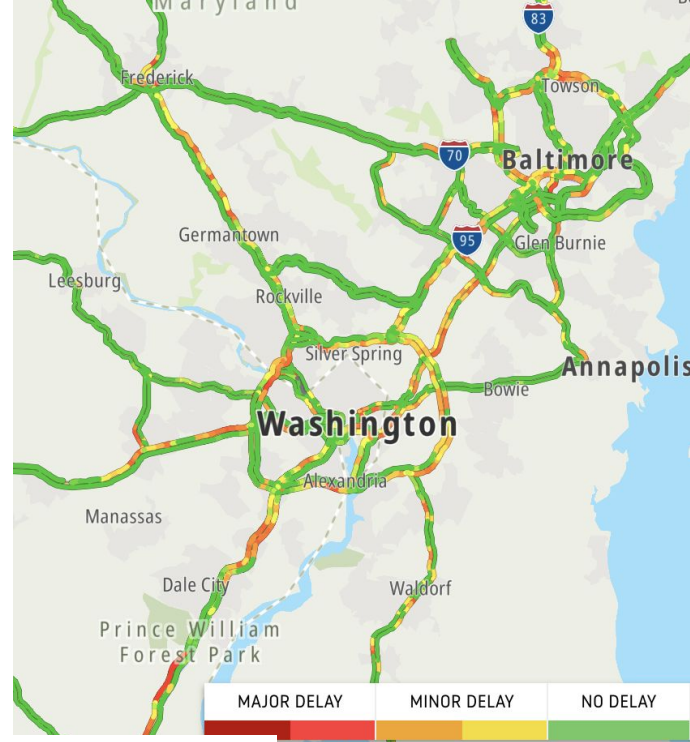
per 30 min trip  
in the evening



## TIME LOST IN RUSH HOUR - PER YEAR

How much extra time was spent driving in rush hours over the year?

**126 hours = 5 days 6 hours**

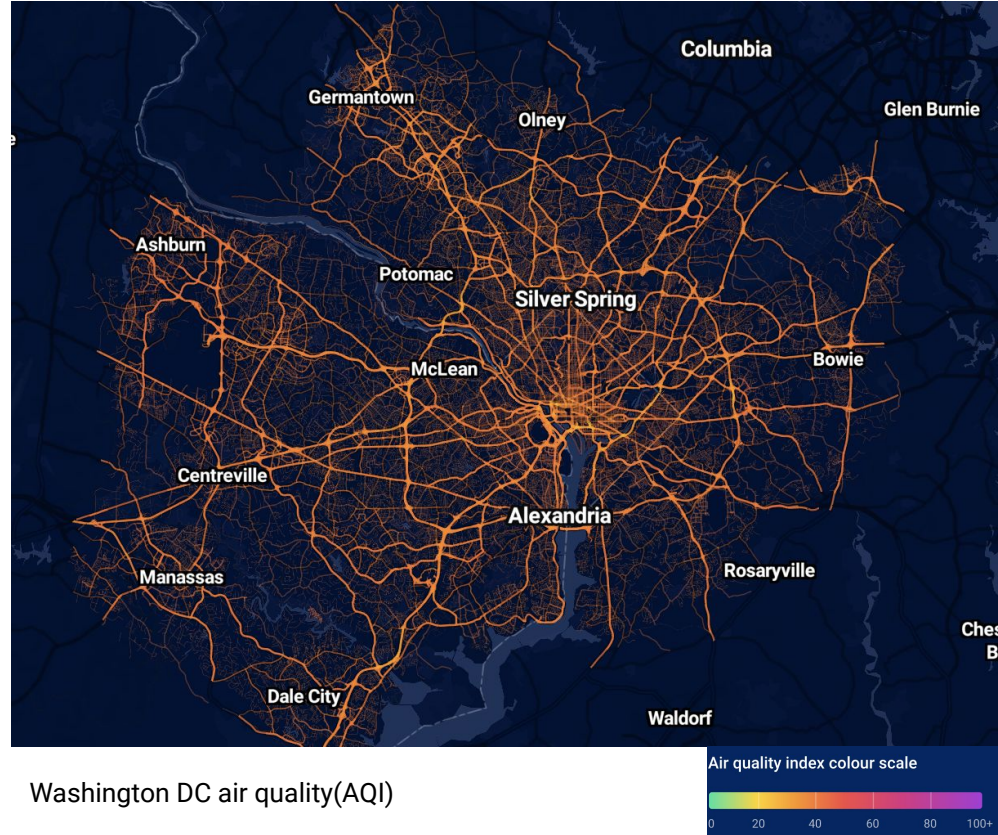




# Traffic congestion leads to

## Air pollution

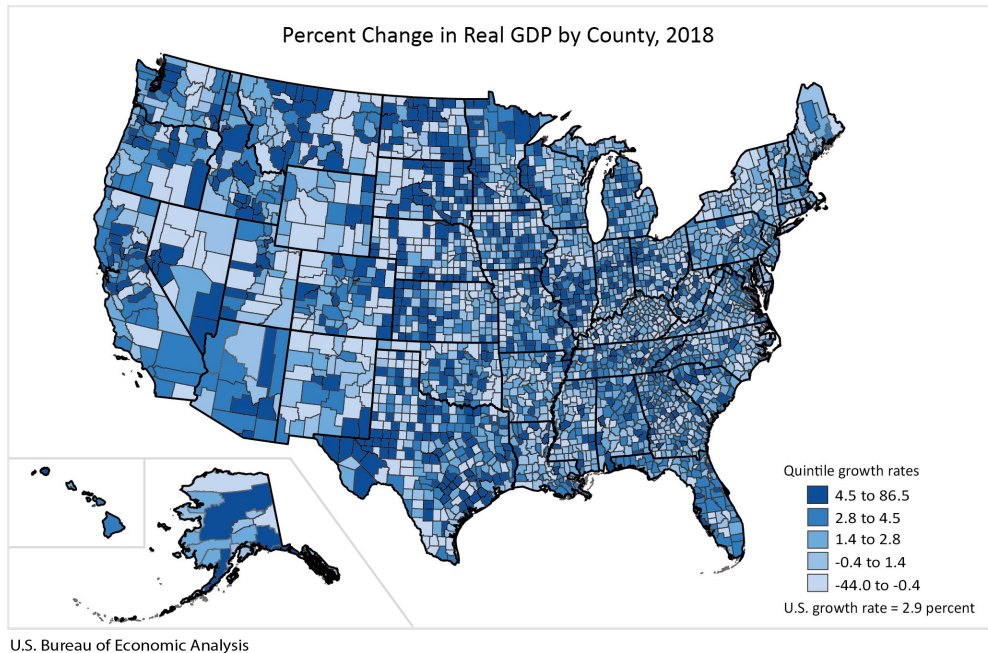
- Vehicle emissions have become the dominant source of air pollutants, including carbon monoxide (CO), carbon dioxide (CO<sub>2</sub>), nitrogen oxides (NO<sub>x</sub>), and particulate matter (PM).



# Traffic congestion leads to

## Economic effects

- In the US, the total cost of lost productivity caused by congestion is \$87 billion in 2019.
- The average DC driver lost 126 hours last year thanks to traffic, which cost the local economy \$4.1 billion<sup>2</sup>.



# Traffic congestion leads to

## Health risks

- Public health cost of mortality attributable to congestion in 83 U.S. cities in was \$31 billion.<sup>1</sup>
- Harvard School of Public Health predicts that by 2020:
  - 1,600 premature deaths
  - \$13 billion in total social costs.



# Metropolitan Washington Council of Governments (MWCOC)

## Regional Traffic Count Dataset

Column Name	Column Description
OBJECTID	Object ID
STATION	Traffic count station ID
COMMENTS	Location description of the station
COUNTY	County
ABPair	Link A-B Pair ID in the TPB road network
Sta_Dir	Direction of the station
Motorcycles	Motorcycle count
PassCars	Passenger car count
Buses	Bus count
SUTrucks	Single unit truck count
CTrucks	Cargo truck count
TOD	Time of day
TotVol	Total volume
StartDate	Start date
EndDate	End date
TimeLabel	Time label
AllTrucks	All truck count
IsPerm	Flag whether the station is permanent
PickupsPanelVans	Pick-ups and panel vans
XCOORD	latitude
YCOORD	longitude



# Exploring the Data gives insights

In this challenge, we are trying to understand the traffic patterns and effects of different external factors on the patterns.

---

# Analysing the car flow

Using the last letter of station direction value to get the car flow direction

Define the angle range of the direction

Traffic flow Visualization, Network Visualization, Statistical analysis



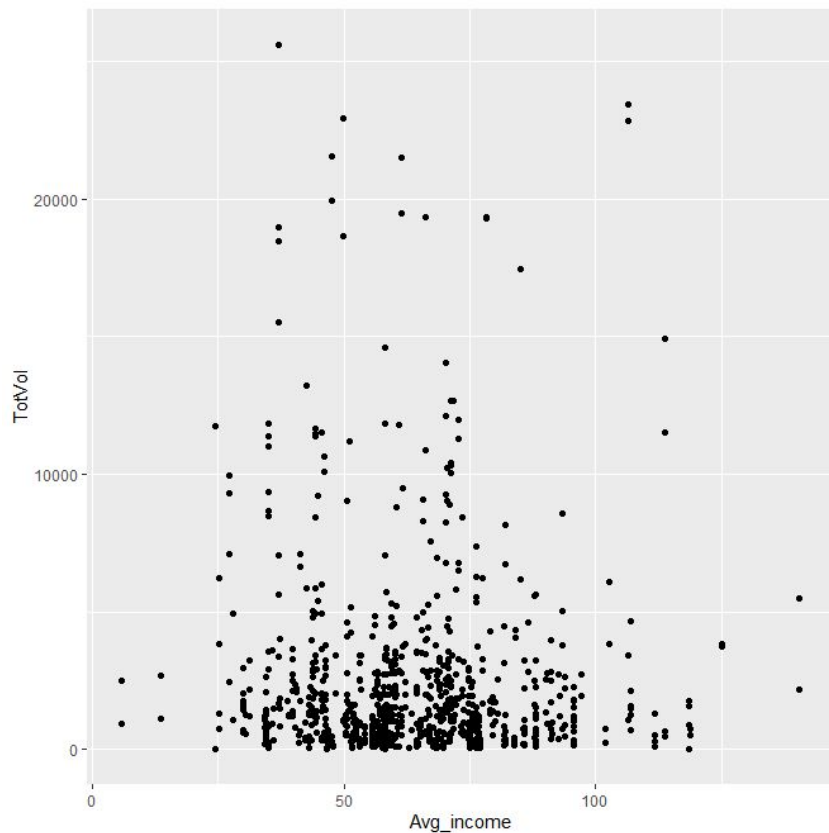
Calculating the relative direction angle by using coordinates and trigonometric function

Using dummy variables to show the results and categorize values

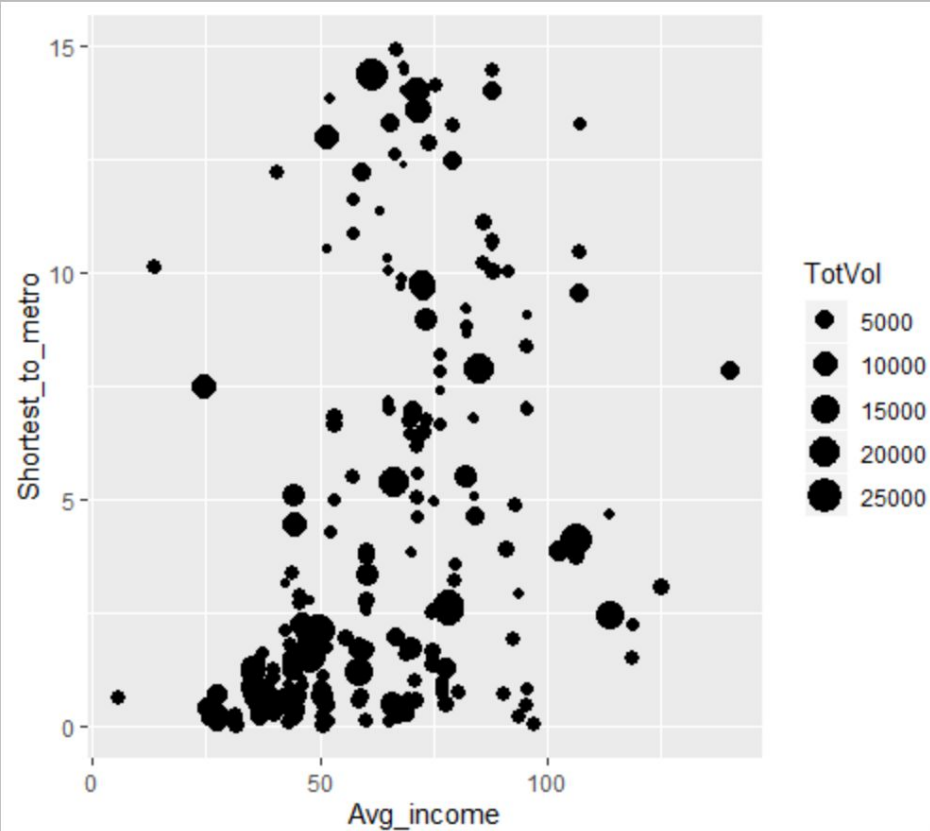
# Using statistical techniques explains our data

We use Poisson regression  
to explain vehicle counts at  
traffic stations by several  
external factors

---



Relationship between **average household income** (ten thousands dollars) and **total traffic volume** (count)



Relationship between **average household income** (thousand dollars), **shortest distance to metro station** (miles) and **total traffic volume** (count)

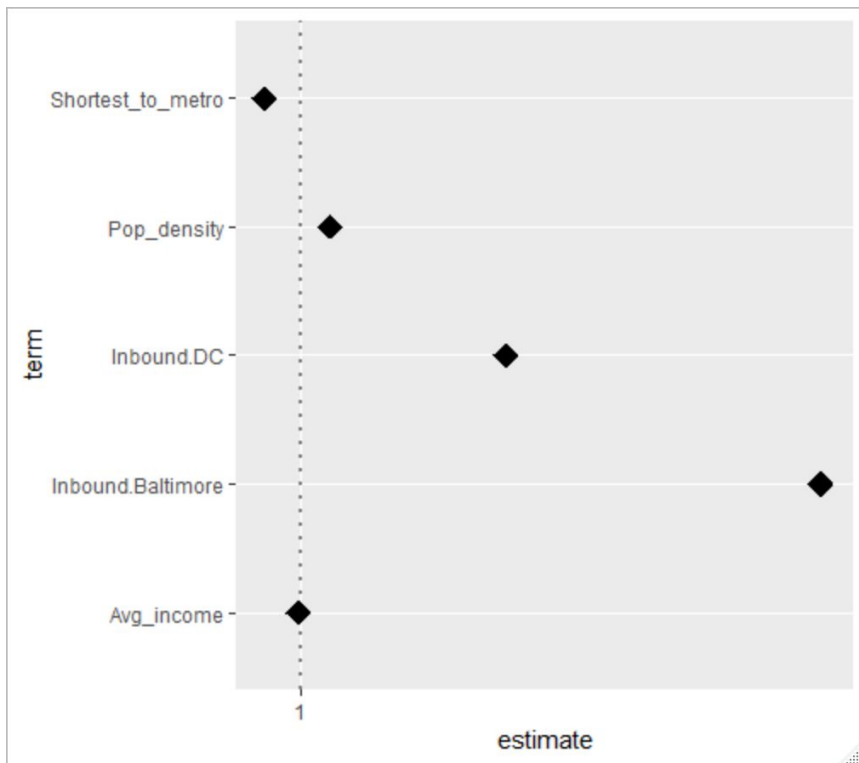


## Here are our predictors:

- ❑ **Average household income (ten thousand dollars)**
- ❑ **Population density (people per square miles)**
- ❑ **Distance to the nearest metro station (miles)**
- ❑ **Direction of the traffic flow**
  - ❑ Inbound DC
  - ❑ Inbound Baltimore
  - ❑ Neither inbound DC nor Baltimore



# Poisson - ONE MODEL



Coefficient plots for Poisson model

## POSITIVE INFLUENCE

Inbound-Baltimore direction ~ Vol **44.92%**↑

Inbound-DC direction ~ Vol **15.78%**↑

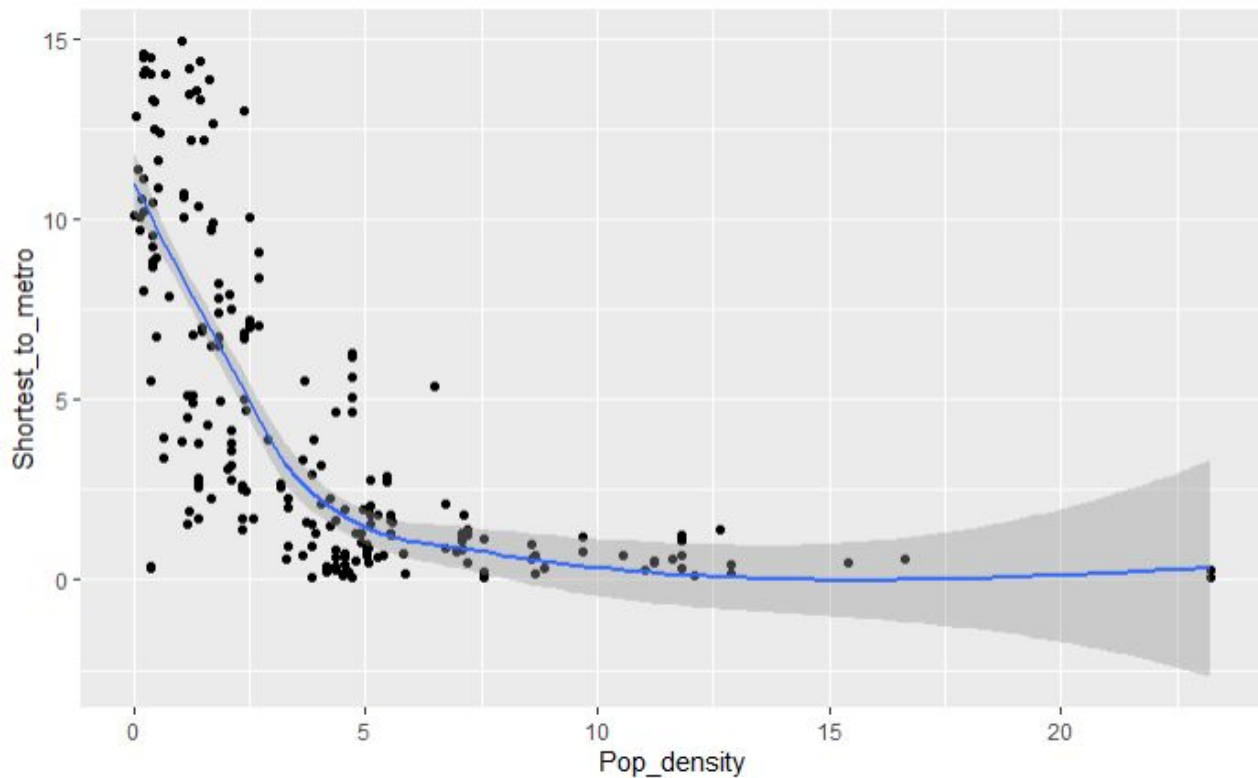
People per square miles **1k**↑ ~ Vol **2.08%** ↑

## NEGATIVE INFLUENCE

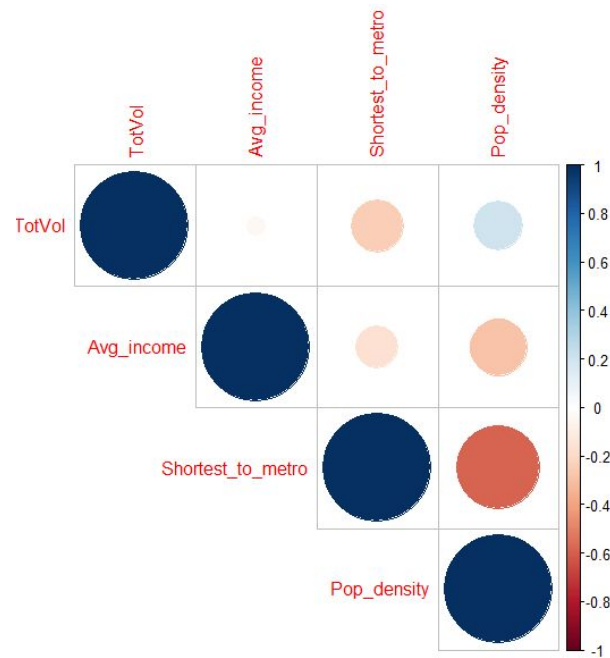
Avg Household income **\$10k**↑ ~ Vol **1.80%**↓

Distance to the nearest metro **1 mile**↑ ~ Vol **2.59%**





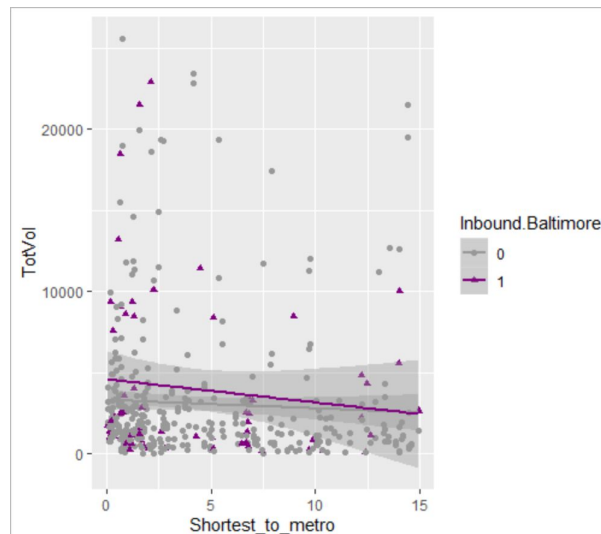
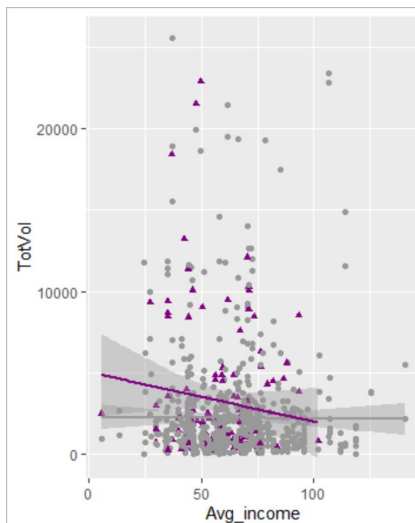
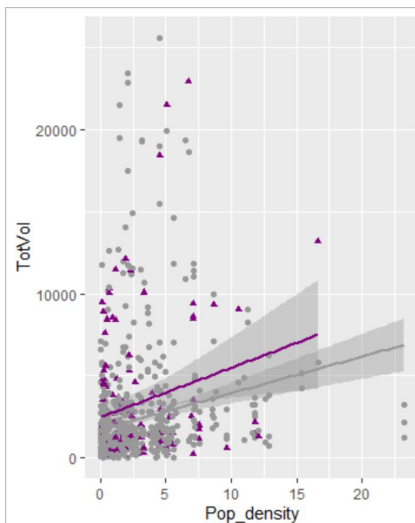
*Relationship between population density (thousand people) and shortest distance to metro station (miles)*



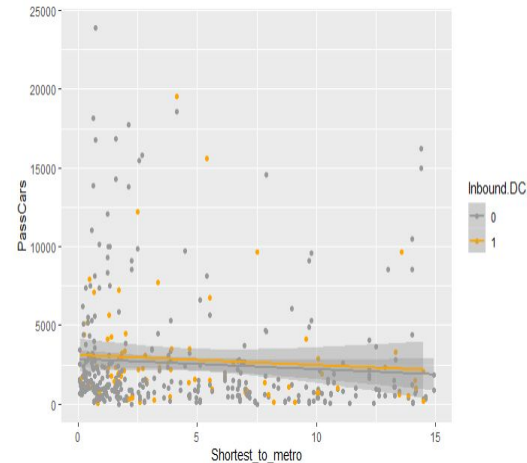
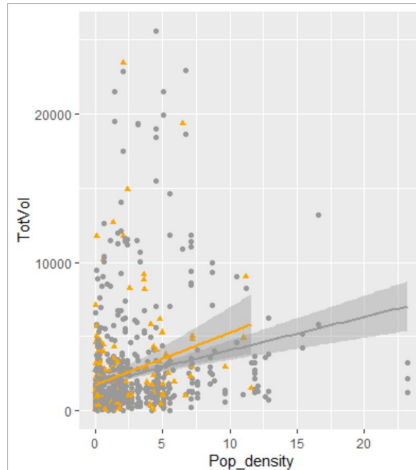
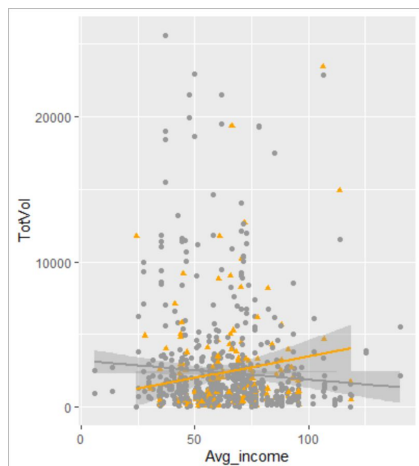
*Correlation matrix*

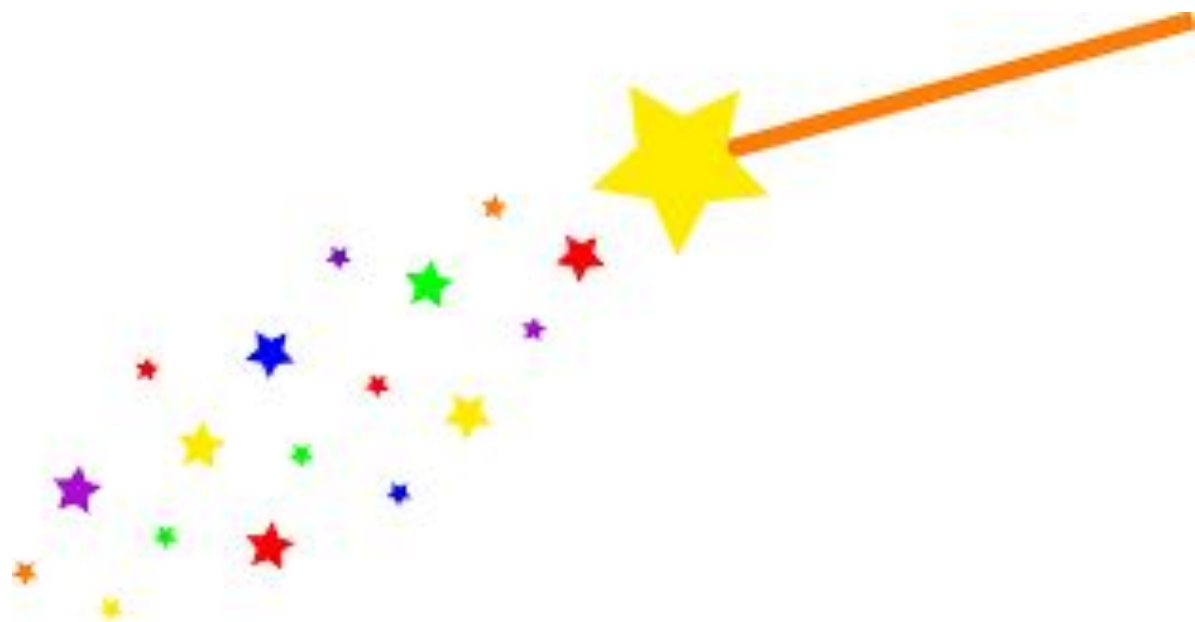


**Direction  
matters a lot.**



**What if we take  
direction of traffic  
flowing as a filter?**







# THREE MODELS

We divide our data into three segments by their directions:

To DC only

To Baltimore only

Neither DC nor Baltimore



# Suggestions

## For City Road

### 1. Tidal lane

According to the flow map, tidal lane should be set more during morning rush hour towards main cities like Washington DC and Baltimore.

### 2. Traffic lights control

To duration of green light at interactions that near most busy traffic station sites.





# Suggestions

## For Highway:

3. Charge more during morning rush hour  
For highways towards Washington D.C. or Baltimore, like the roads we mentioned in the car flow graphes, we can charging more fees to reduce traffic.
4. Add carpool lane



# References:

- 1) <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC4243514/>
- 2) <https://www.weforum.org/agenda/2019/03/traffic-congestion-cost-the-us-economy-nearly-87-billion-in-2018/>
- 3) [https://www.tomtom.com/en\\_gb/traffic-index/washington-traffic](https://www.tomtom.com/en_gb/traffic-index/washington-traffic)