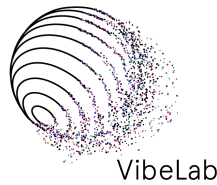




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Understanding the Impact of the Nightlife Industry on Urban Mobility in NYC and Charting COVID-19 Related Disruption

Kaifu Ren (kr2516), Nicholas LiCalzi (nl949)

Yingyuan Zhang (yz6378), Yutong Zhu (yz6080)

Sponsor: VibeLab

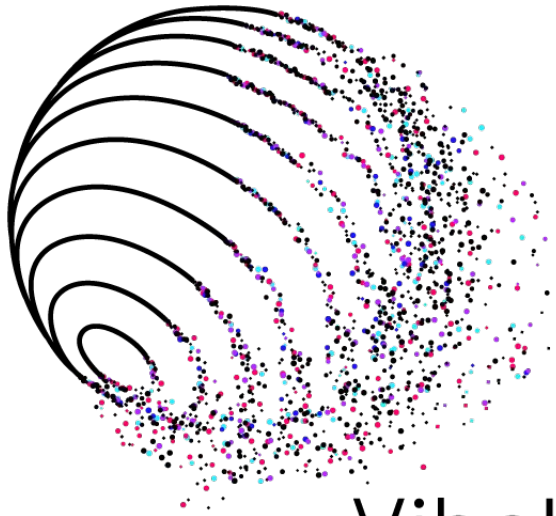
CUSP Mentor: Kim Mahler

July 23 2020



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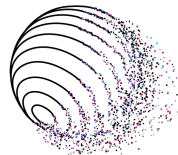
VibeLab





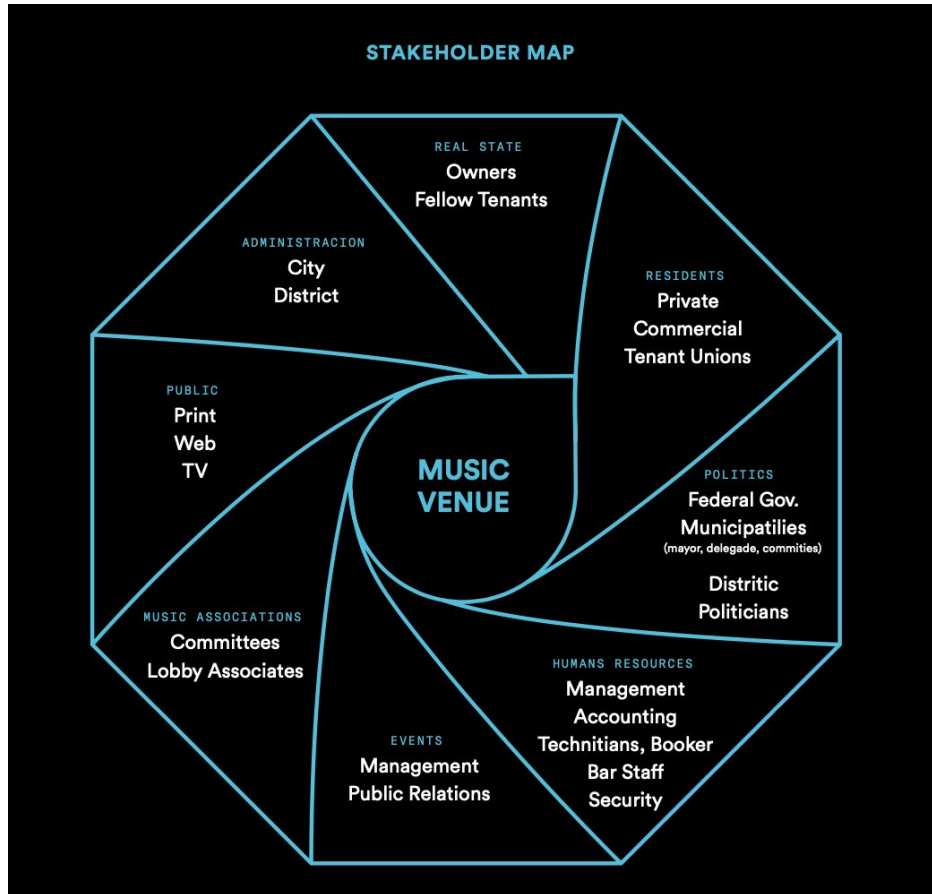
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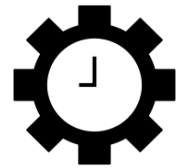
Creative Footprint



Source: Creative Footprint,
“Measuring Live Music Space in NYC” (2018)

Why is this project important?

NYC's Music Industry accounts for:



57,000 Jobs



\$4.7B in Wages



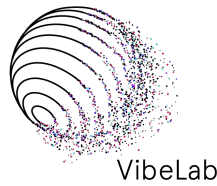
\$21B in Economic Output

Source: Mayor's Office of Media & Entertainment,
"Music in New York City" (2017)



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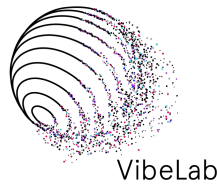
Research Questions

- **Where does NYC's nightlife occur?**
 - **Is it centralized or diffuse?**
 - **Are there specific (or surprising) clusters?**



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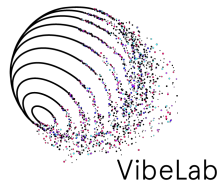
Research Questions

- Where does NYC's nightlife occur?
 - Is it centralized or diffuse?
 - Are there specific (or surprising) clusters?
- How do people access nightlife in NYC?
 - Are there any discernible predictors of difference in mode choice?



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Research Questions

- Where does NYC's nightlife occur?
 - Is it centralized or diffuse?
 - Are there specific (or surprising) clusters?
- How do people access nightlife in NYC?
 - Are there any discernible predictors of difference in mode choice?
- **What changes do we see in citywide/boro-wide nighttime transportation ridership due to COVID-19?**

We built a baseline model of winter day and nighttime transportation with data from 2019, expecting to see significant variation given the arrival of the COVID-19 pandemic and related disruptions.

Establish a **Baseline**:

"Normal" transit conditions (same period in 2019)

Capture all night time transportation data



Source: nycgo.com



Transition Period:

Pre-COVID closures,
number of cases surging
but businesses still open

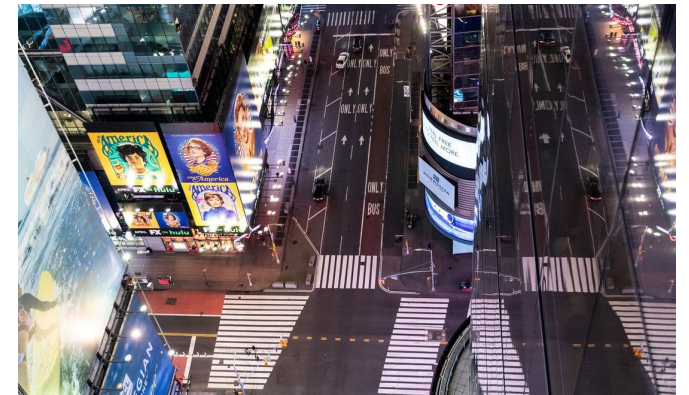
- Did venue clusters see a decline in traffic before they were ordered shut?
- Was impact greater than in other areas?



Current **Snapshot**:

Post-COVID outbreaks (2020)

Changes in mobility and activity once
nightlife businesses/venues were closed



Source: Business Insider



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Data

- **Venue Data:**

Creative Footprint (494 venues) + Google Maps (3,601 venues)

- **Transportation Data:**

TLC (For-hired vehicles) + MTA + Citibike

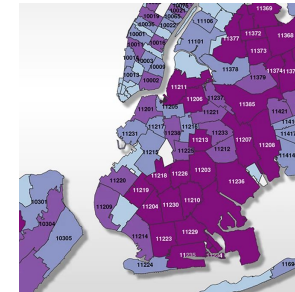
- **Demographics:**

Land Scan: day and night population

- **Built Environment Features:** PLUTO (NYC DCP)

- **Weather:** Weather Underground

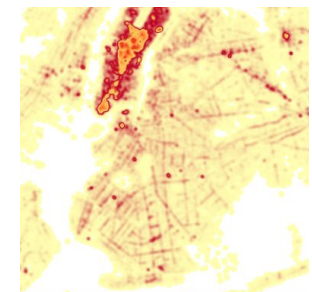
Zip Codes



Taxi Zones



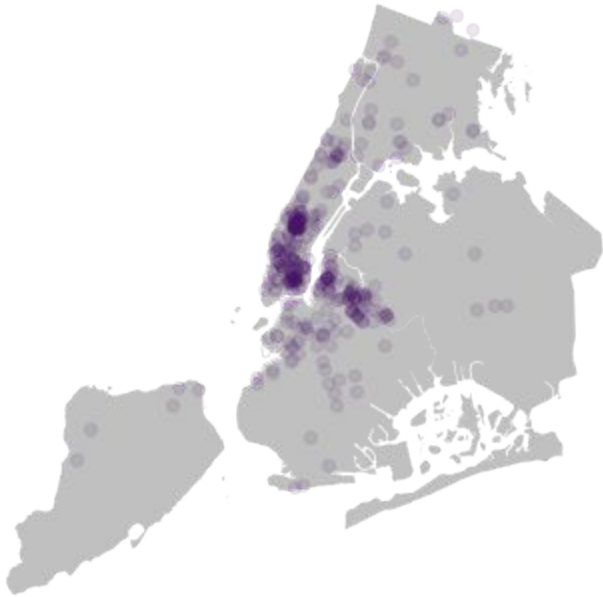
Census Blocks



Nightlife Venues

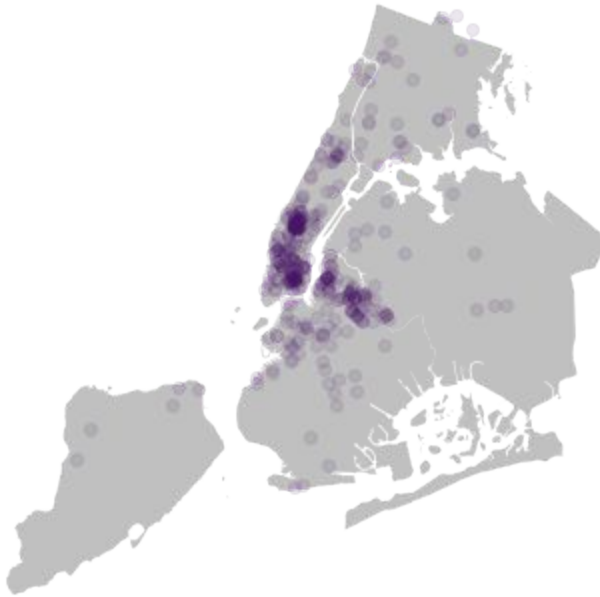
Venue Data (Creative Footprint survey)

Venues Unique to Creative Footprint Survey
Total: 400 nightlife venues

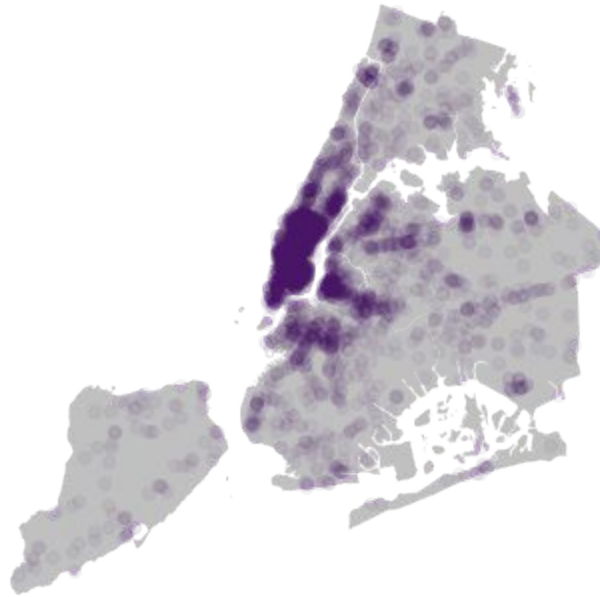


Venue Data (supplementing with Google)

Venues Unique to Creative Footprint Survey
Total: 400 nightlife venues



Google Maps Nightlife Venues
Total: 3601 bars and clubs





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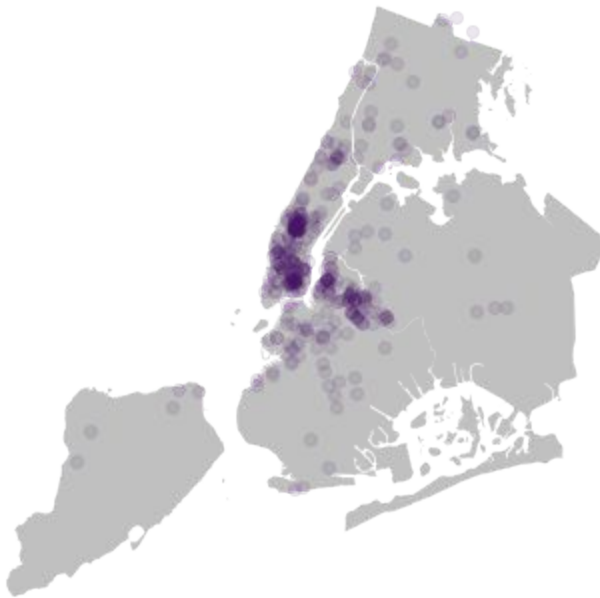
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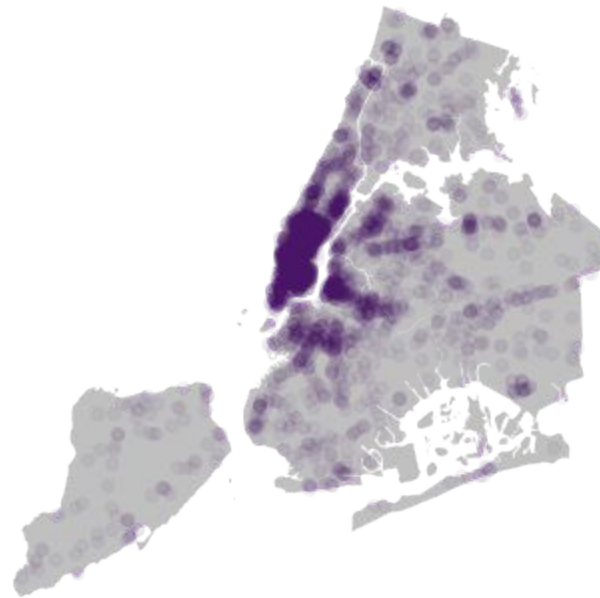
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Aggregated Venue Data

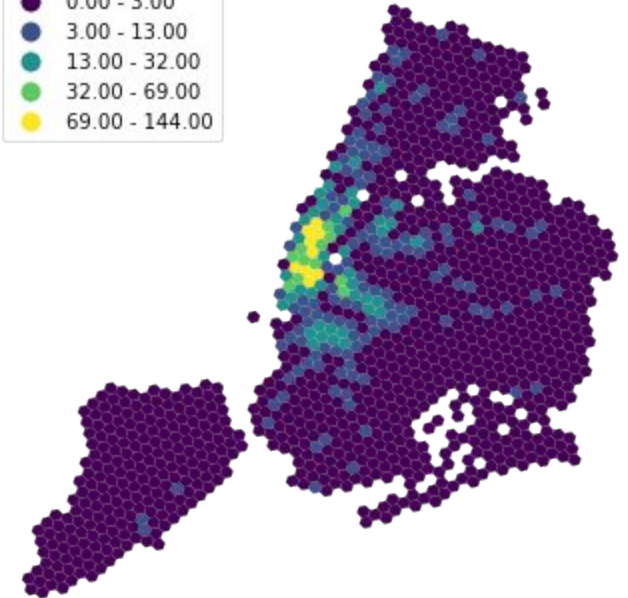
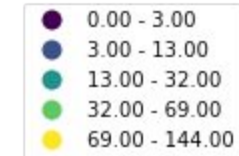
Venues Unique to Creative Footprint Survey
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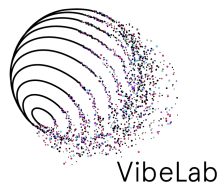
Venues aggregated to hexbin scheme
Total hexes: 1199





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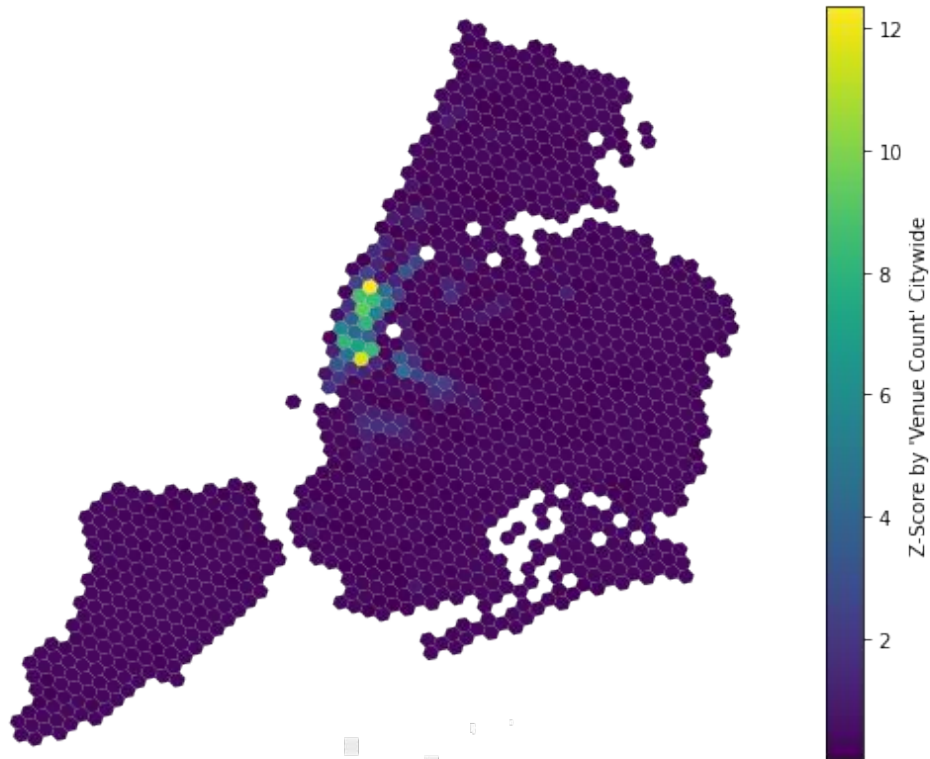
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Detecting Venue Clusters: The Wrong Way

Citywide Venue Count
Z Scores by Hex



$\mu = 3$ venues
 $\sigma = 11$ venues



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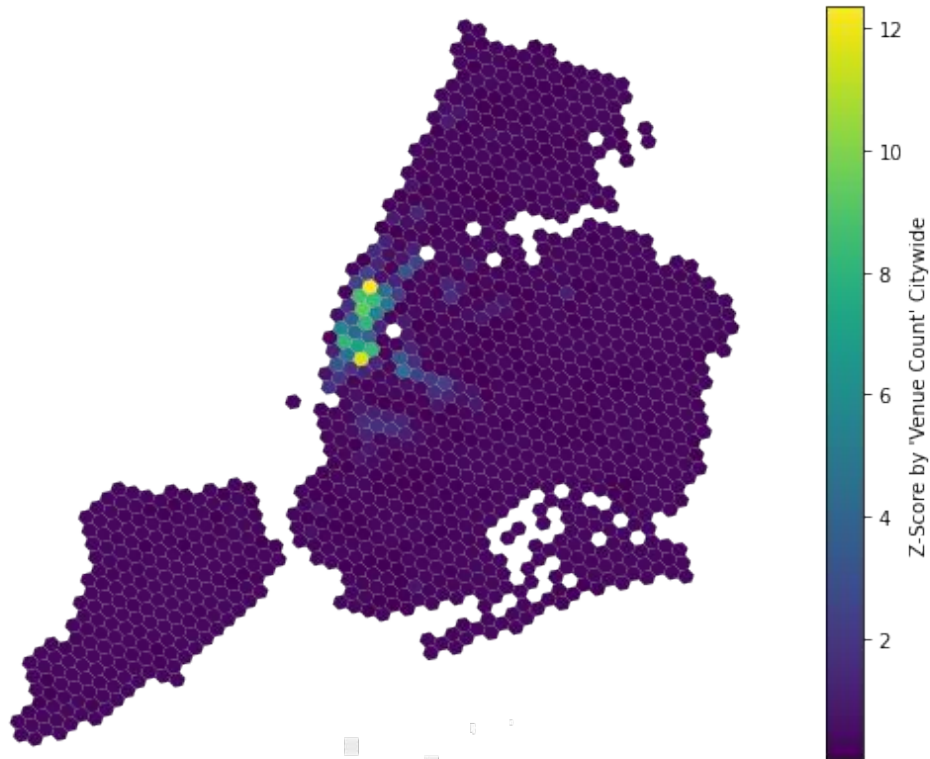
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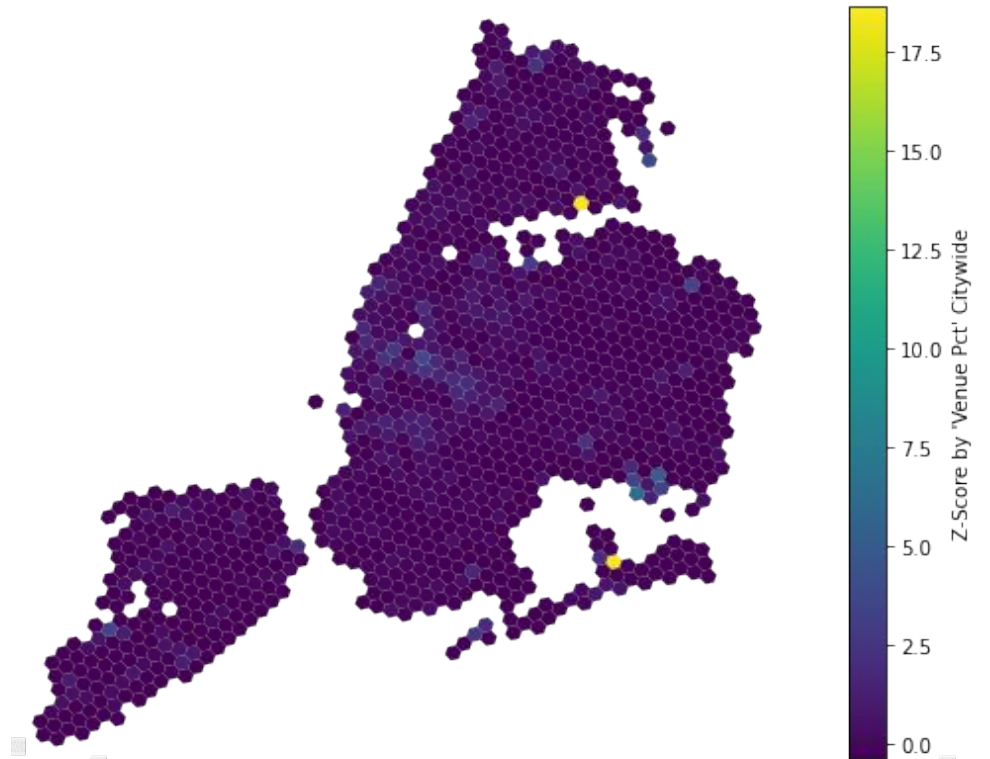
Detecting Venue Clusters: The Wrong Way

Citywide Venue Count
Z Scores by Hex



$\mu = 3$ venues
 $\sigma = 11$ venues

Citywide Venue Share of Businesses (%)
Z Scores by Hex



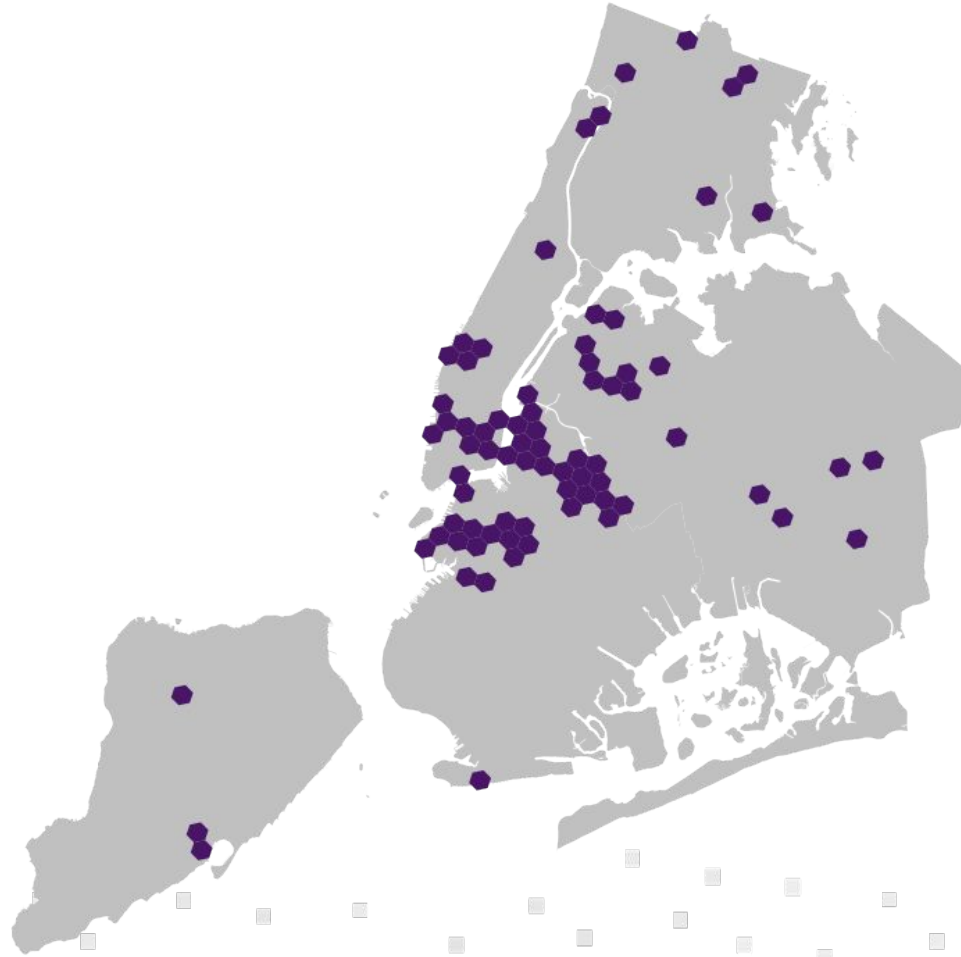
$\mu = 0.66\%$ venue share
 $\sigma = 1.76\%$ venue share



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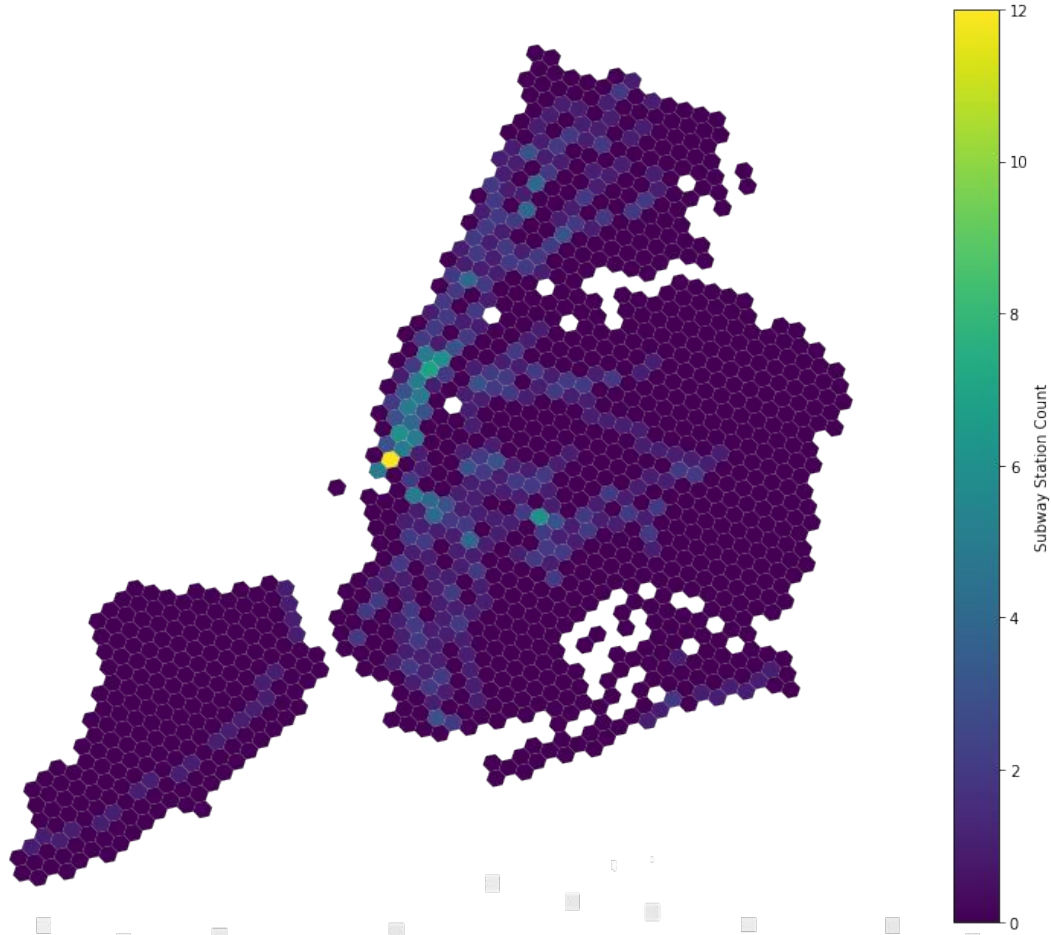
Citywide Nightlife Clusters
w/ Venue Share Z-score > 0.5 & Min. 150 Businesses

Detecting Venue Clusters: The Right Way

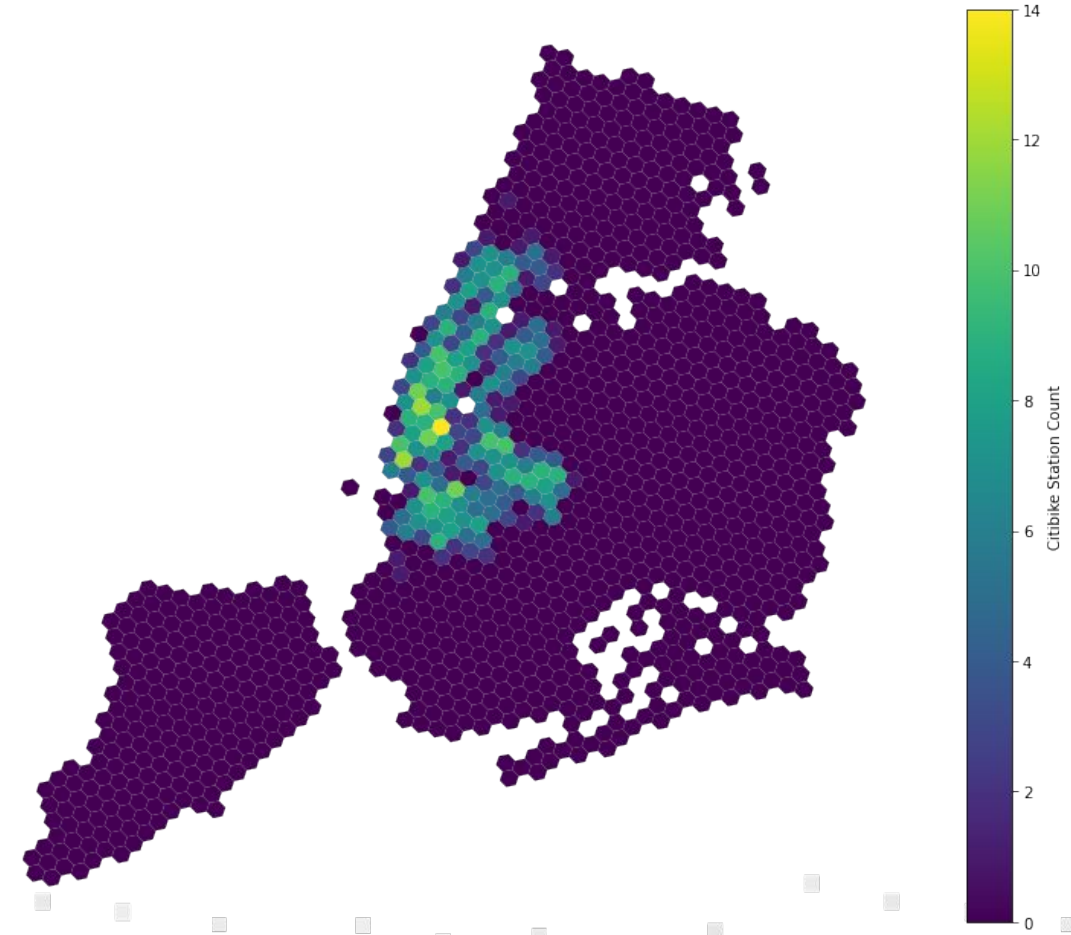


Transportation Data (Public Transit)

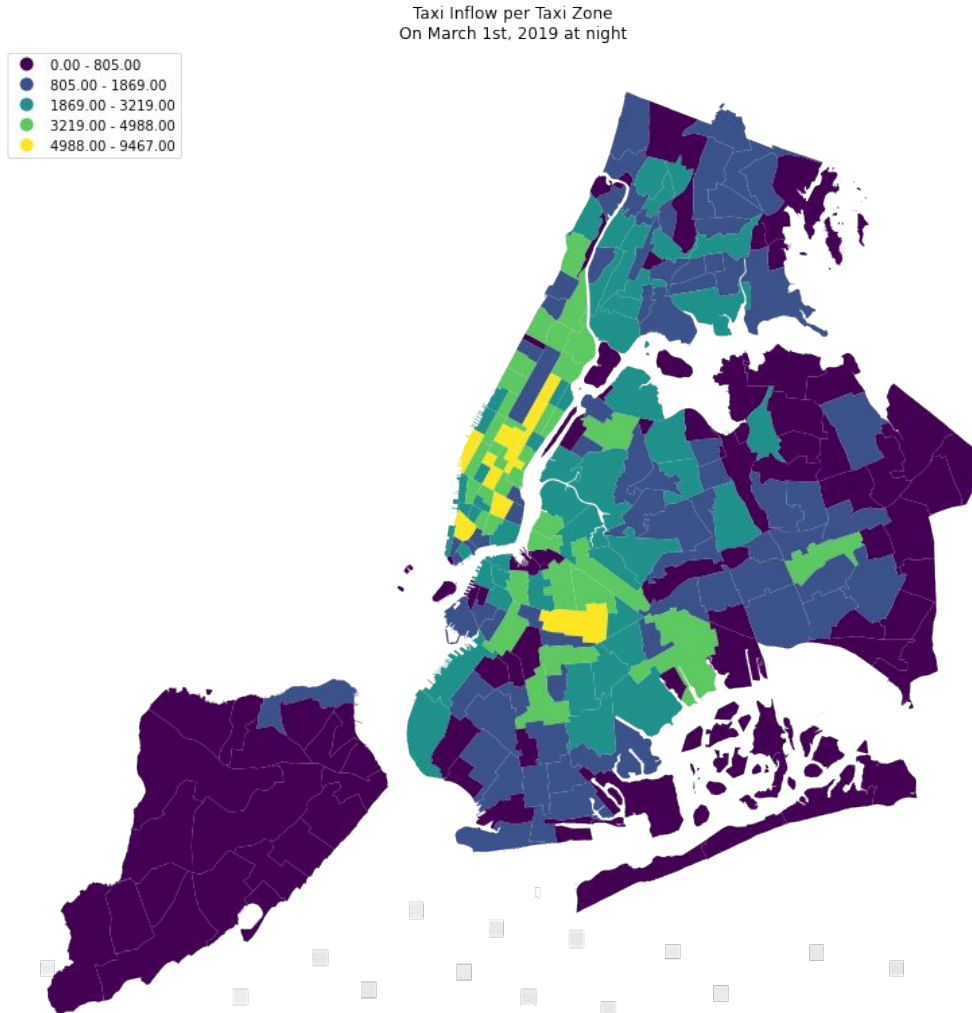
Distribution of Subway Stations across NYC



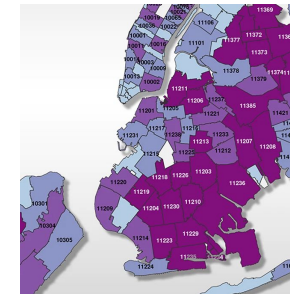
Distribution of Citibike Stations across NYC



Transportation Data (For-Hire Vehicles)



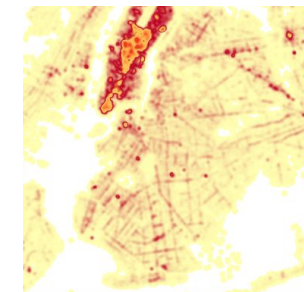
Zip Codes



Taxi Zones



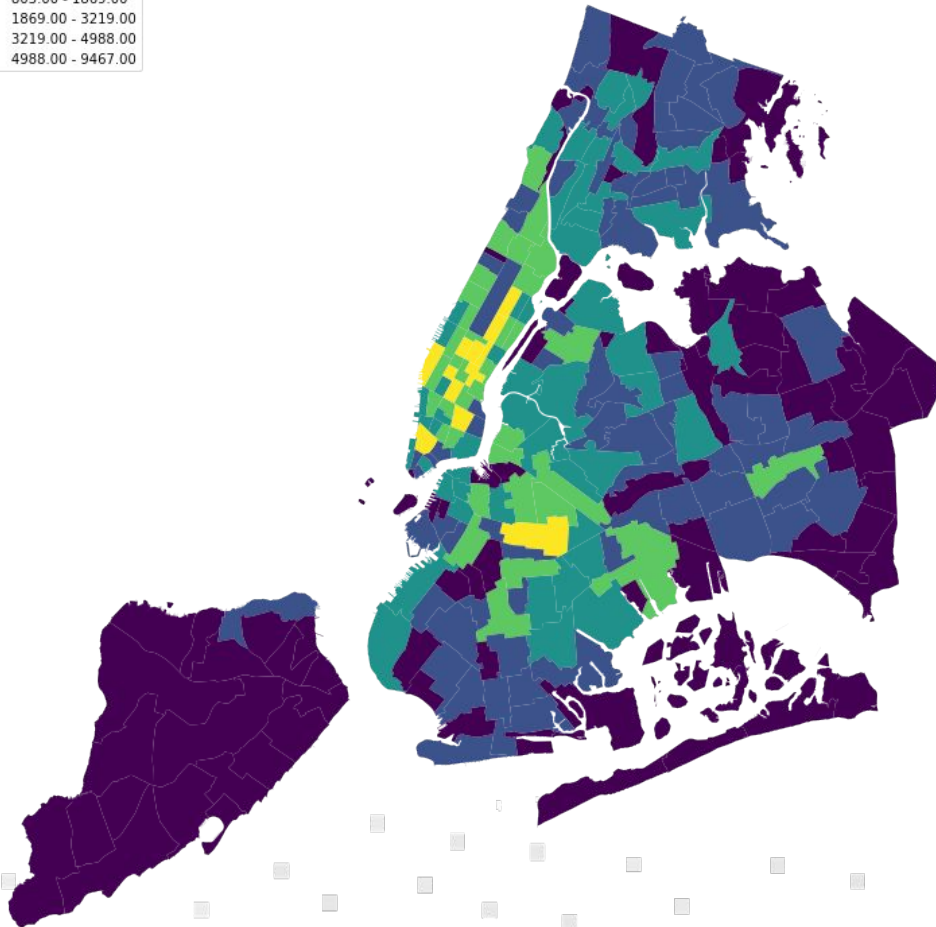
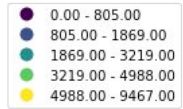
Census Blocks



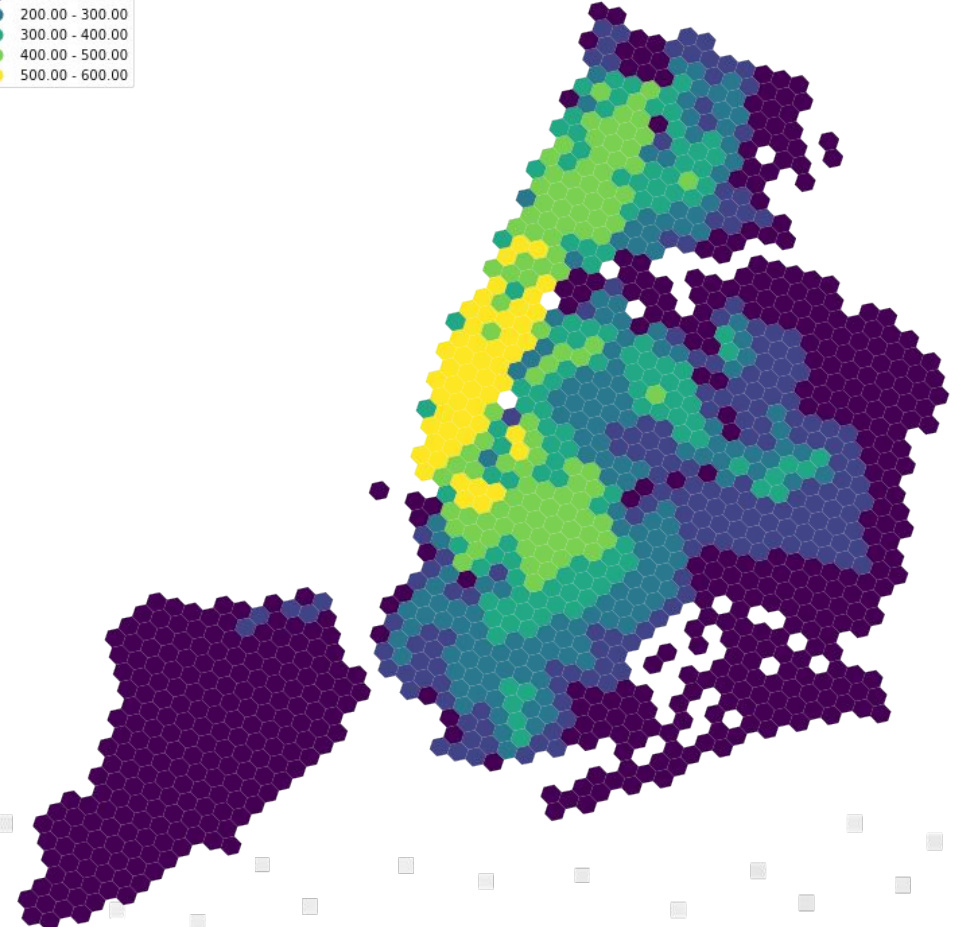
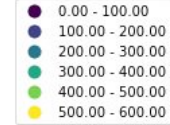
Nightlife Venues

Transportation Data (For-Hire Vehicles) Areal Interpolation

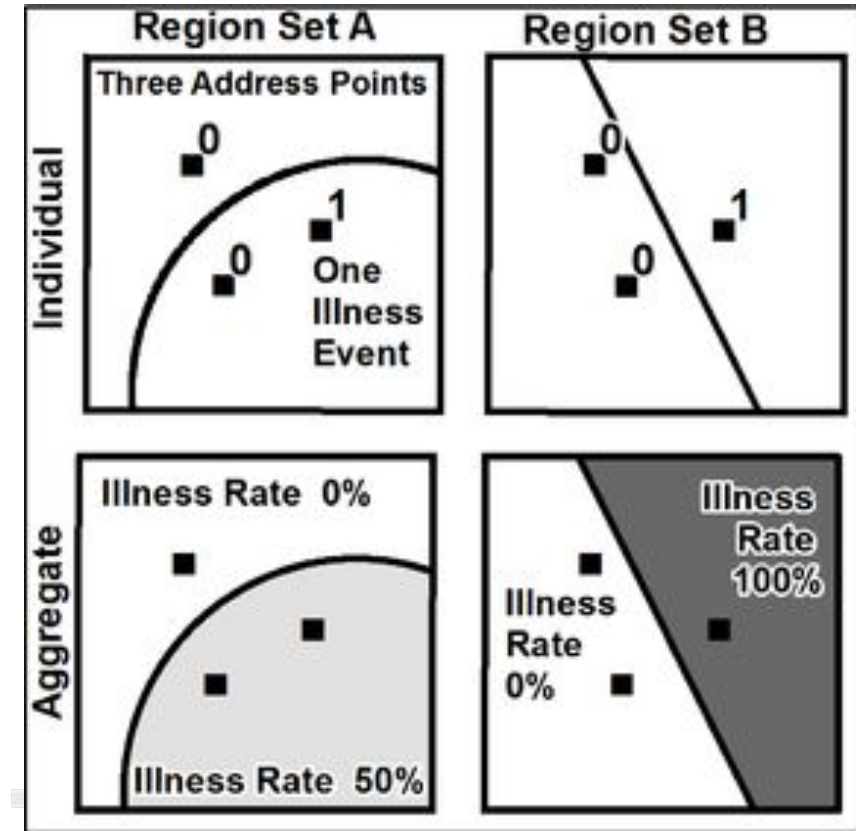
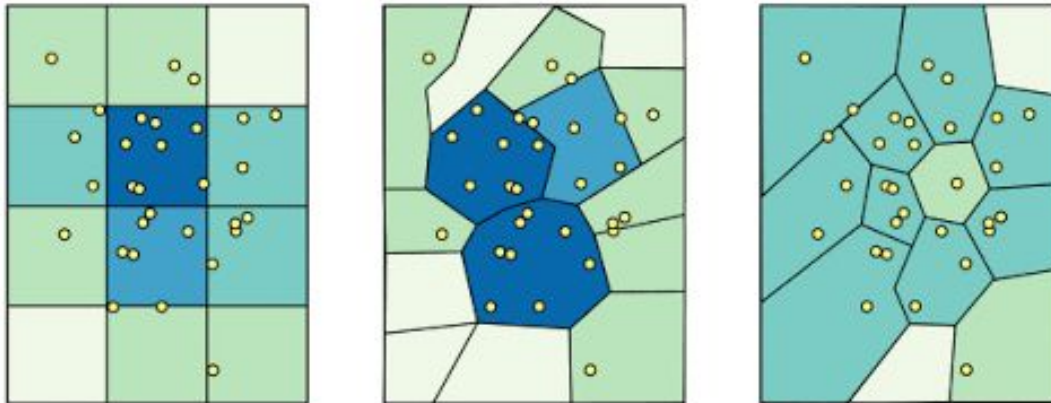
Taxi Inflow per Taxi Zone
On March 1st, 2019 at night



Taxi Inflow per Hexbin
On March 1st, 2019 at night

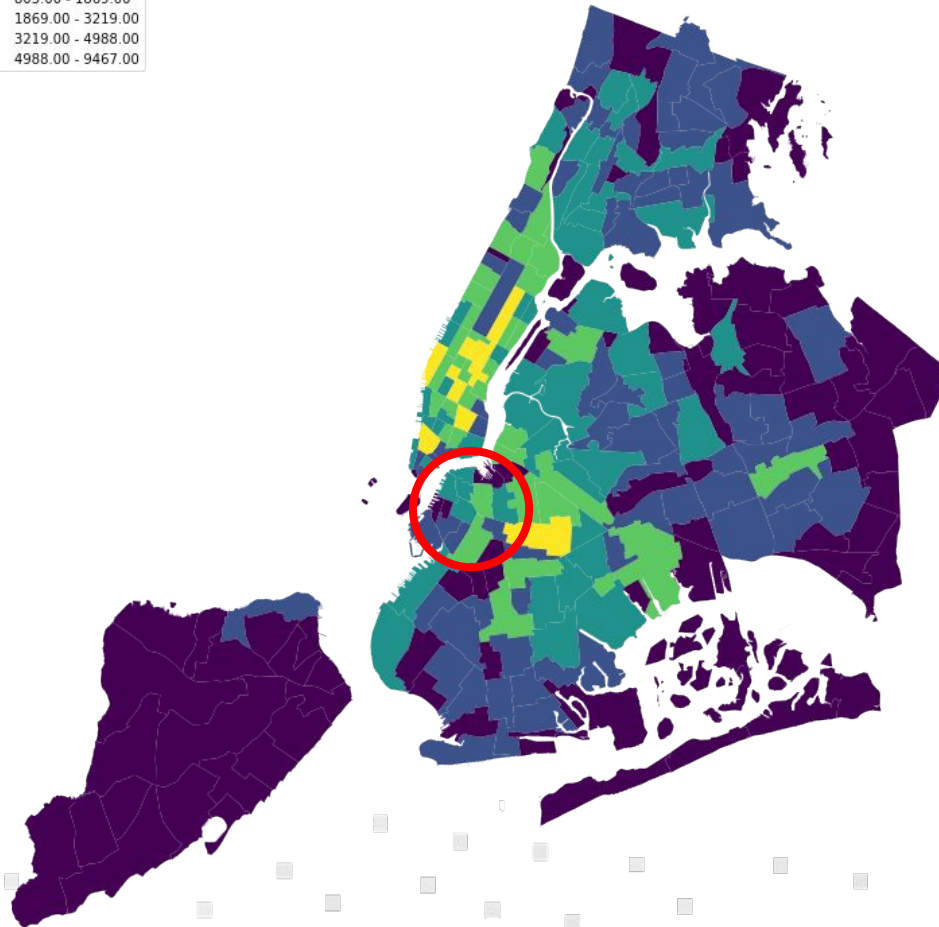
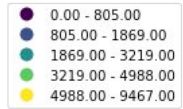


The Modifiable Areal Unit Problem (MAUP)

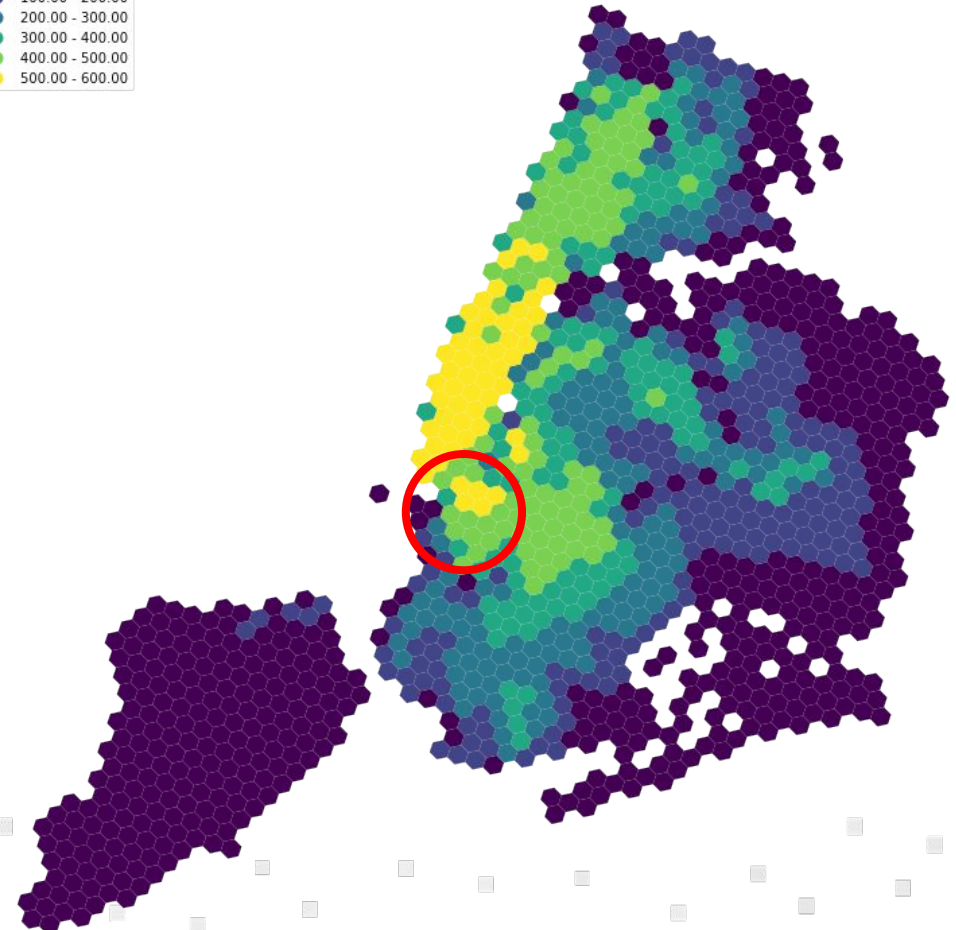
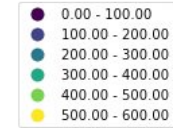


The Modifiable Areal Unit Problem (MAUP)

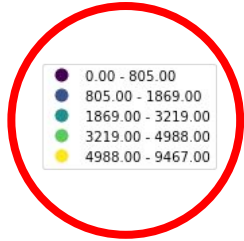
Taxi Inflow per Taxi Zone
On March 1st, 2019 at night



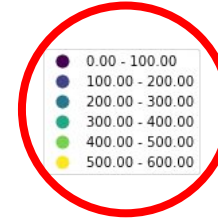
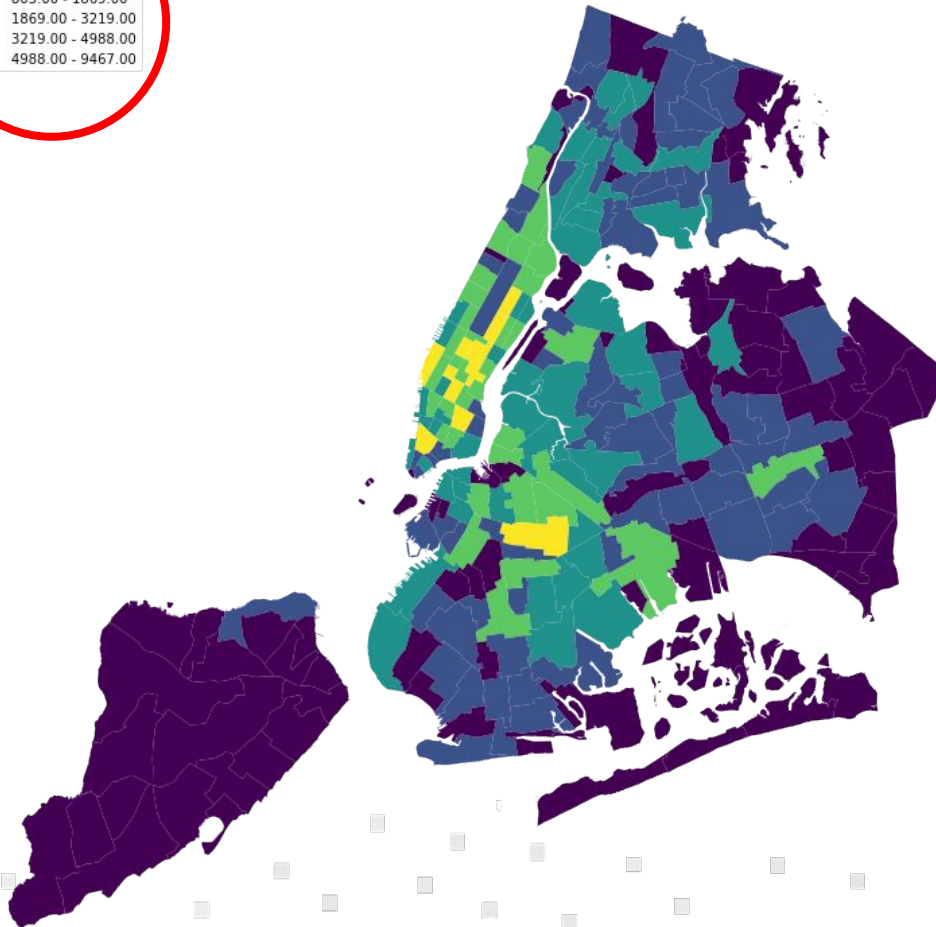
Taxi Inflow per Hexbin
On March 1st, 2019 at night



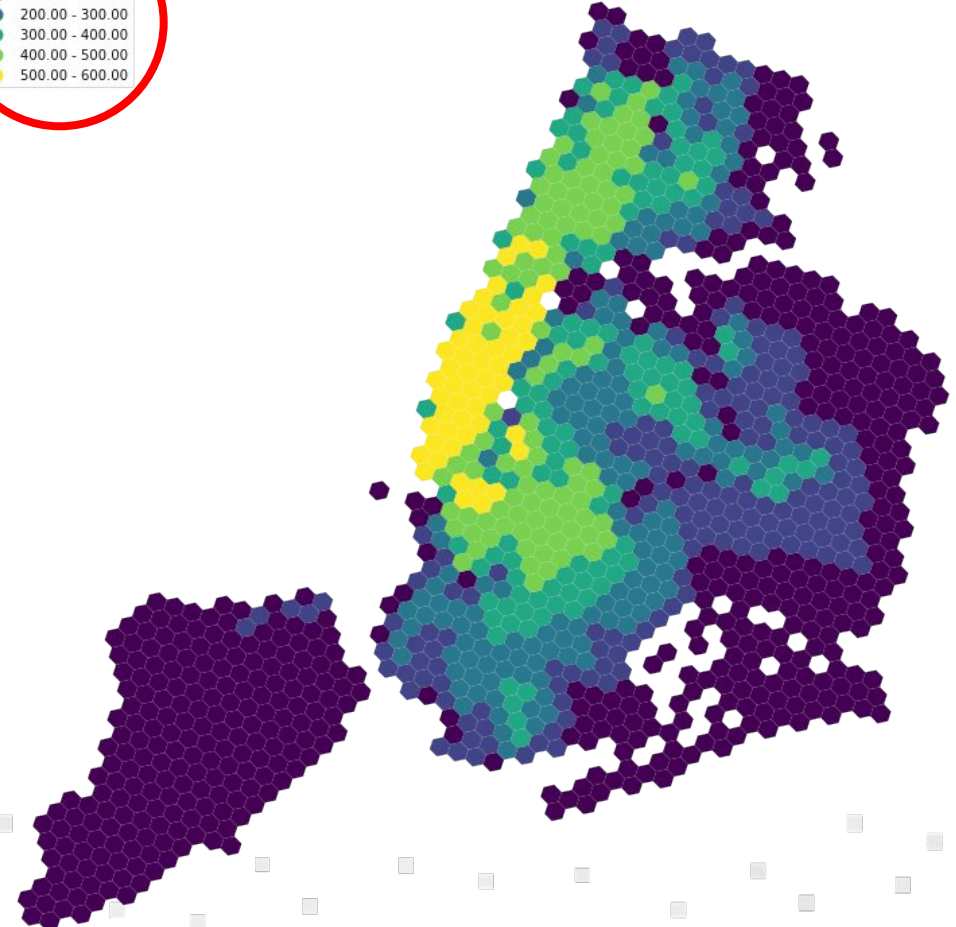
The Modifiable Areal Unit Problem (MAUP)



Taxi Inflow per Taxi Zone
On March 1st, 2019 at night



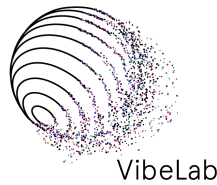
Taxi Inflow per Hexbin
On March 1st, 2019 at night





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Method - Data Engineering

Big Data

The sheer size of trip records

Streaming

Process data row by row; no monopolizing RAM

Final Dataset Preparation

Aggregating to **daily day and night** data intervals ranging from **Jan. 1 - Apr. 30** for the **years 2019 and 2020**

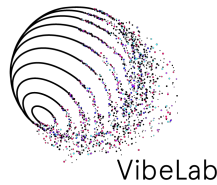
Hex ID	Day/ Month/ Year	Night?	FHV/ Citibike/ MTA Inflow/ Outflow	Night/ Day Population	Venue/ Business Count	Res/ Comm Area	Temperatur e/ Rainfall	Covid?	Boro Code	Nos. of Citibike/ MTA stations
str	timestamp	bool	int	int	int	float	float	bool	int	int

Available via Github and available for public use and research (<https://github.com/nlicalzi/cuspcapstone>).



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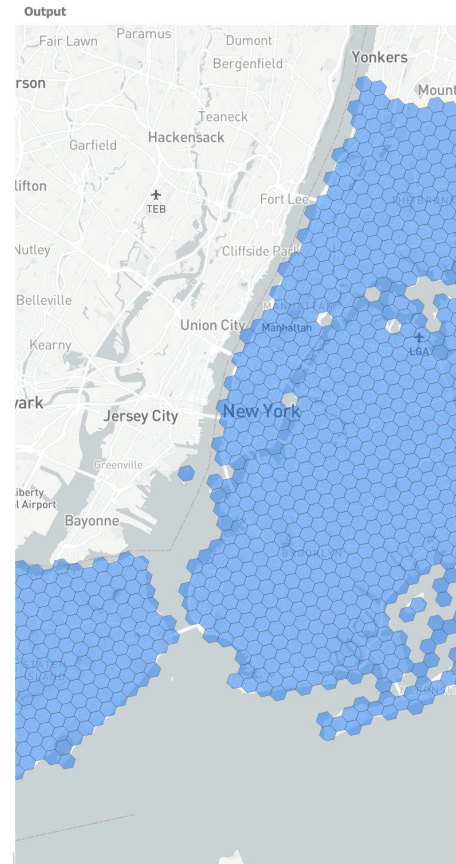
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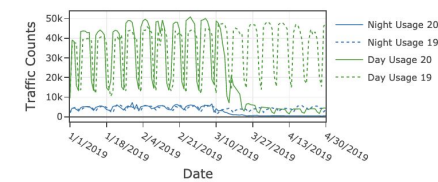
Method - Data Visualization

Data Visualization Interface

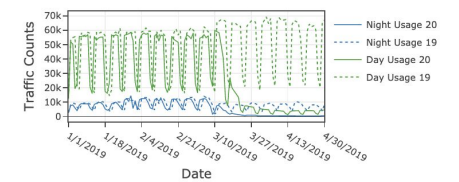
- D3.js and the MapboxGL Javascript API
- Monitor daily or nightly changes of different mobility usage in hexagons
- Time series analysis
- Comparison to baseline model



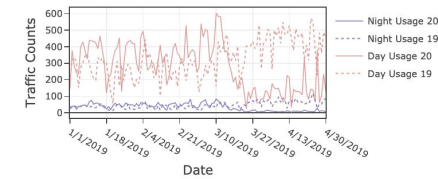
MTA Outflow Usage 2019 vs 2020



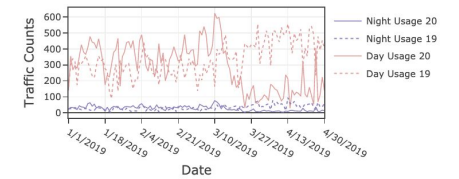
MTA Inflow Usage 2019 vs 2020



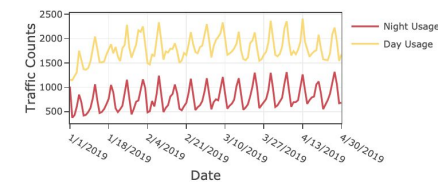
Citibike Outflow Usage 2019 vs 2020



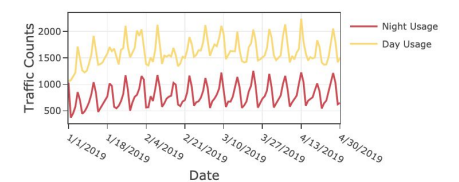
Citibike Inflow Usage 2019 vs 2020



FHV Outflow Usage 2019



FHV Inflow Usage 2019



Method - Regression Analysis

$$DailyInflow_i = \beta_0 + \beta_1 DayoftheWeek_i + \beta_2 AveTemp_i + \beta_3 Precipitation_i + \beta_4 NightDummy + \varepsilon_i$$

$$VenueCounts_i = \beta_0 + \beta_1 BuseinssCount_i + \beta_2 TMTAInflows_i + \beta_3 CitiBikeInflow_i + \beta_4 FHVInflows_i + \beta_5 Stationcounts_i + \beta_6 boroDummy + \varepsilon_i$$

$$NightDailyInflow_i = \beta_0 + \beta_1 StationCount_i + \beta_2 BusinessCount_i + \beta_3 LotsArea_i + \beta_4 ResidentialArea_i + \beta_5 OfficeArea_i + \beta_6 RetailArea_i + \beta_7 NightPopulation_i + \beta_8 BoroughDummy + \varepsilon_i$$

Transportation: MTA, Citibike, Fhv(Yellow Cab, Ubers etc.), Station counts(MTA & Citi)

Year : 2019 and 2020 (Covid-19)

Environment built Features:

Business Counts, Residential, Commercial, Retail Area

Demographic feature: Population data

Weather: Temperature, Precipitation

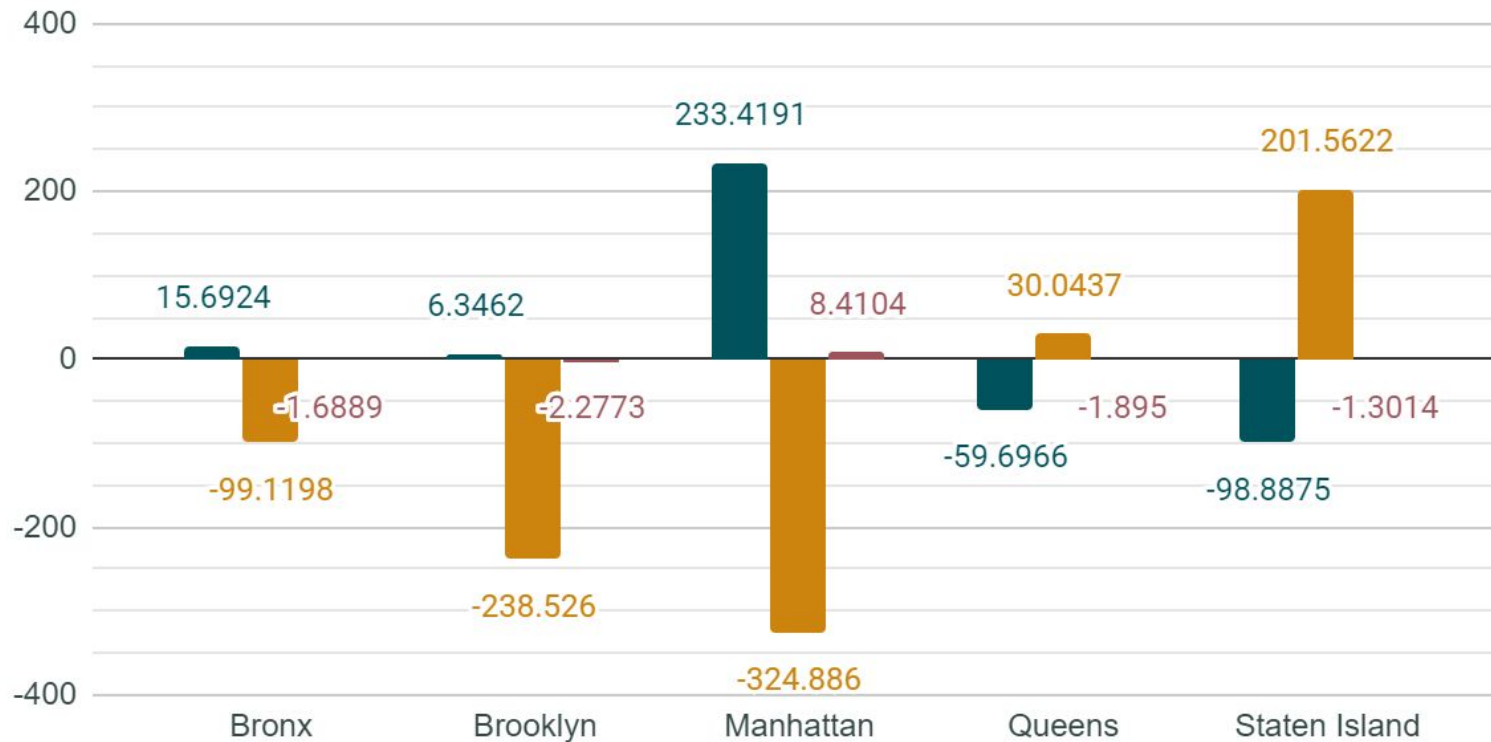
Time-related: Day of the Week, day or night

Result

How do people access nightlife in NYC and are there any predictors of difference in mode choice?

NightTime Inflow Regression based on Borough

FHV MTA Citibike



Unit approximately $\frac{1}{8}$ Central Park.

Nightlife transportation preference

- Staten Island - MTA Preferred
- Manhattan - Taxi Preferred

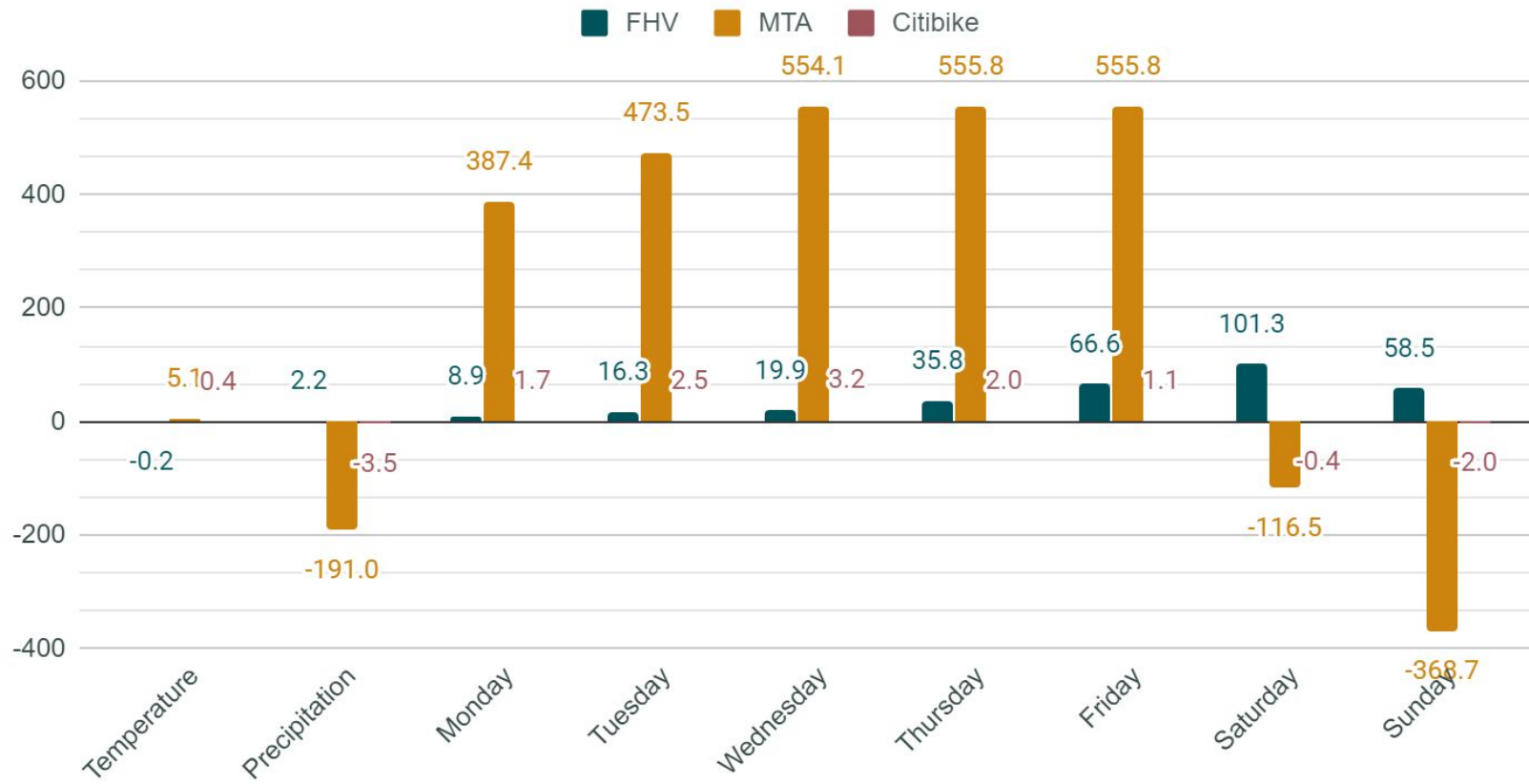
Citi Bike Influence

- Manhattan - More dock locations

Result

What is the relationship between transport mode choice and weather, weekday and weekend and day or night?

Daily Inflow v.s. weather and time-related factors



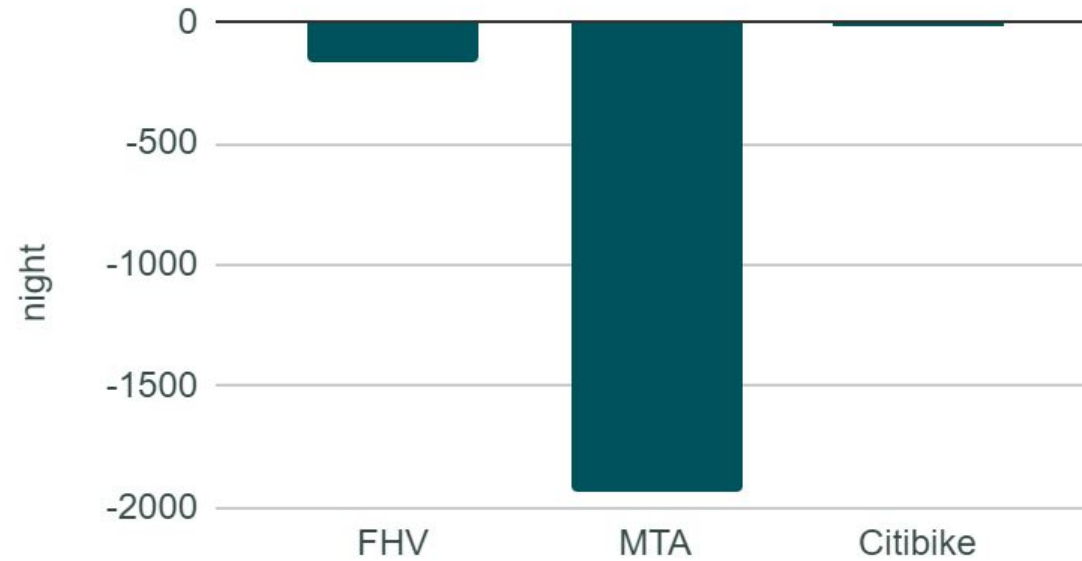
Area size: approximately 1/5 Central Park

- Weekend
 - Taxi > Citi Bike > MTA
- Weekday
 - MTA > Taxi > Citi Bike
- Rainy Day solution
 - Taxi ↑
 - MTA and Citi Bike ↓

Result

What is the relationship between transport mode choice and weather, weekday and weekend and day or night?

night



Area size: approximately $\frac{1}{5}$ Central Park

All Decrease at Night



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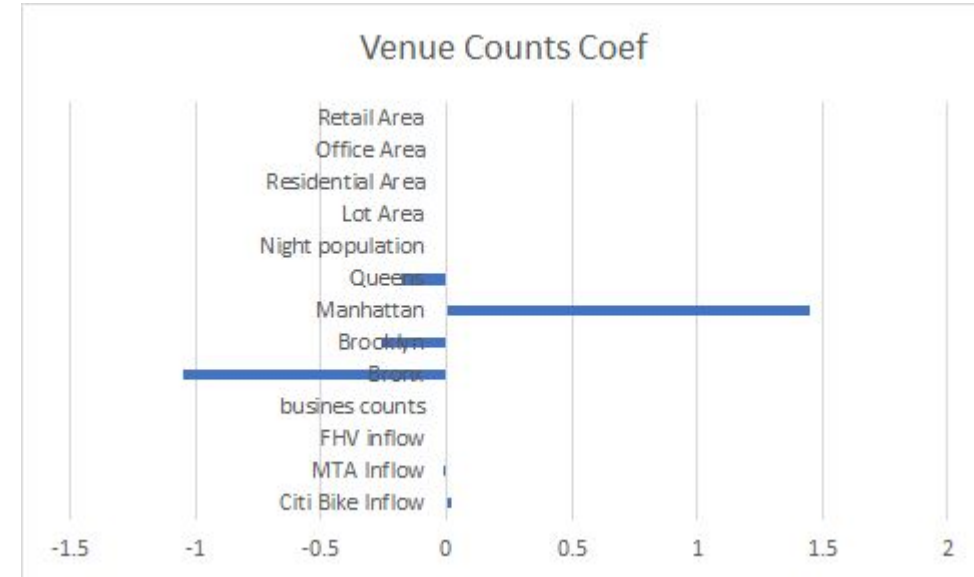
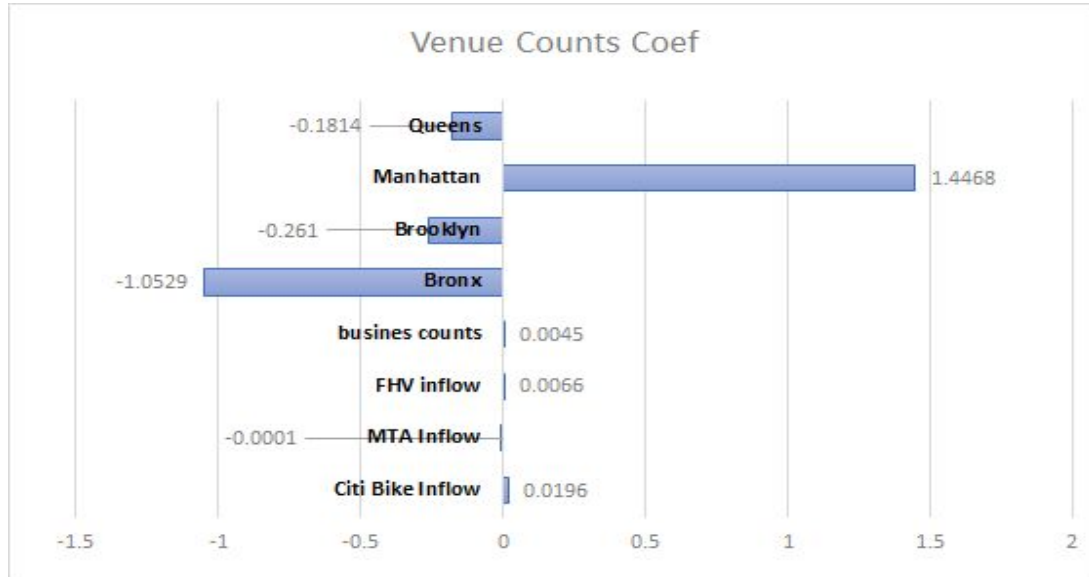
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Result

How does the nightlife venue grows?



- Manhattan has the highest venue counts
- Citibike has a higher positive coefficient on venue counts.
 - traffic congestion at night time
- Different land use has less impact on venue counts
- Venue locations have strong relationships with boroughs
- Little relationship with populations

Result

What is the citywide and borough-wide decrease in transportation utilization due to Covid-19?

Period 1 Feb.12 to 28
Normal Condition

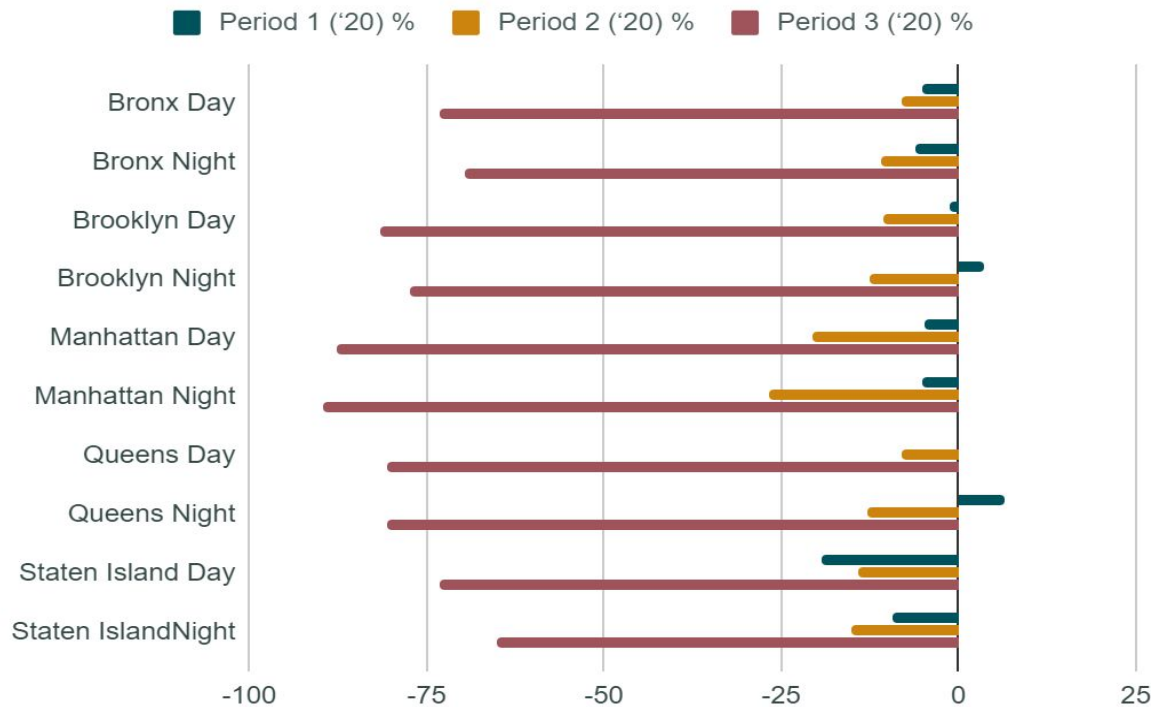


Period 2 Mar.1 to 17
Aware of Covid-19 Outbreak

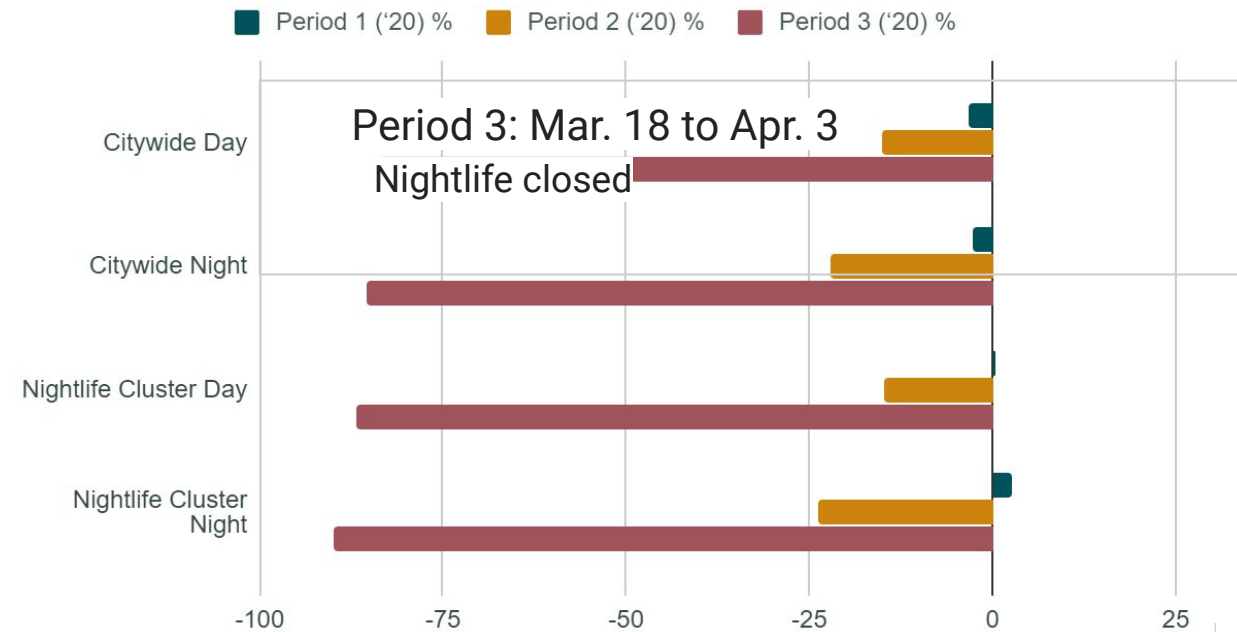


Period 3 Mar.18 to Apr.3
Nightlife Closed

Subway Usage Change 2019 vs 2020



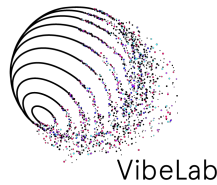
MTA usage change 2019 vs 2020





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Result

What is the citywide and borough-wide decrease in transportation utilization due to Covid-19?

Period 1 Feb.12 to 28
Normal Condition



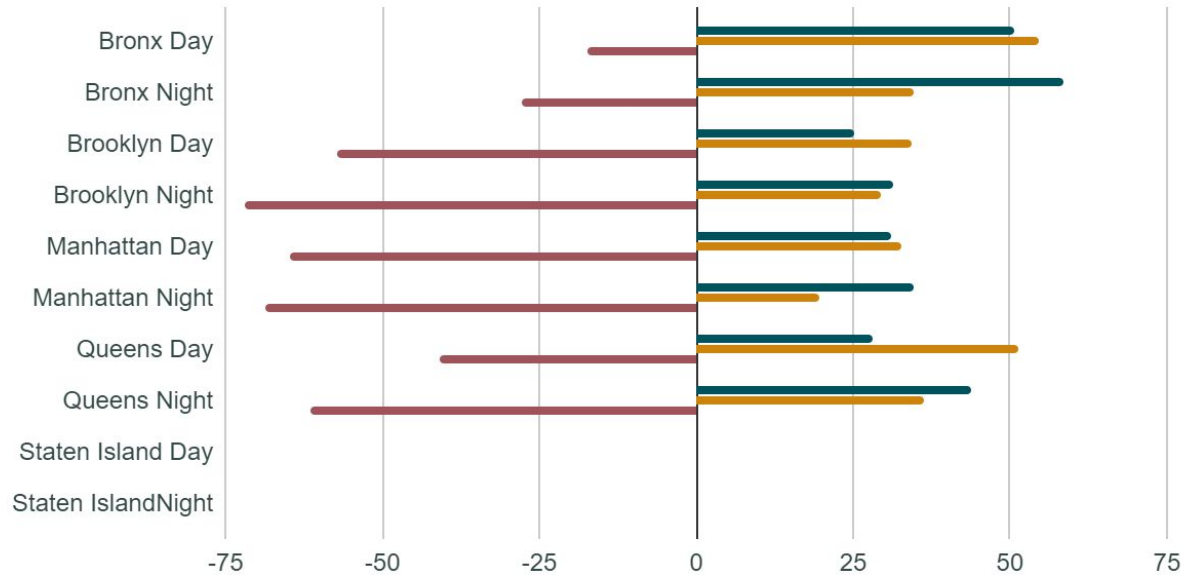
Period 2 Mar.1 to 17
Aware of Covid-19 Outbreak



Period 3 Mar.18 to Apr.3
Nightlife Closed

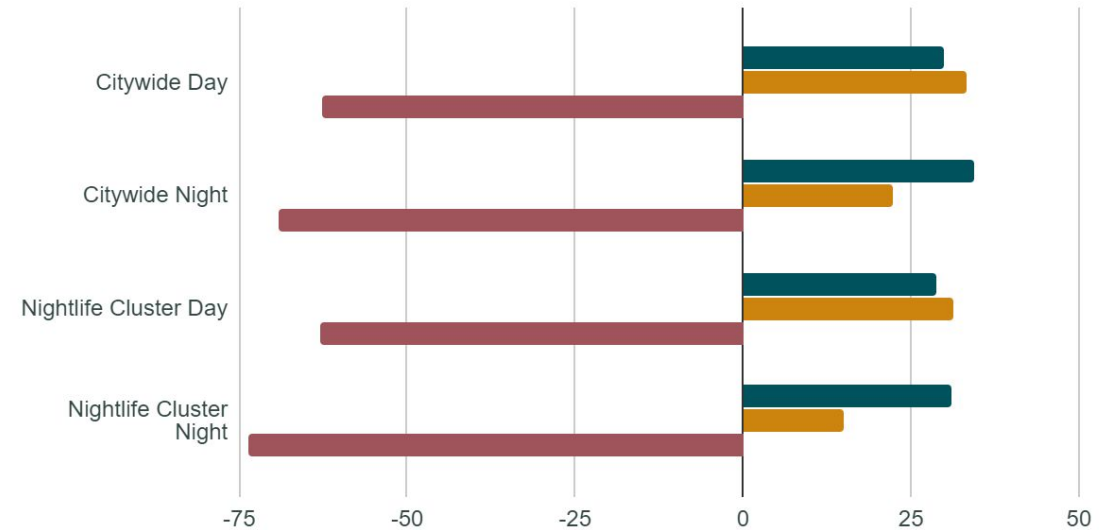
Citi Bike usage change 2019 vs 2020

Period 1 ('20) % Period 2 ('20) % Period 3 ('20) %



Citi Bike usage change 2019 vs 2020

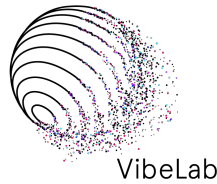
Period 1 ('20) % Period 2 ('20) % Period 3 ('20) %





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Conclusion & Future work

We have outlined a process for identifying nightlife hotspots, combine and analyze disparate datasets to attempt to explain ridership phenomena.

Potential extensions of our work might include:

1. using measures of local/global spatial autocorrelation to identify clusters of nightlife activity
2. the use of Ridge/LASSO regressions to neutralize any collinearity in our datasets
3. placing a greater emphasis on last mile transit options

The COVID-19 pandemic has had devastating effects on the City as a whole, but particularly on the nightlife sector, and we hope that the research and processes contained in this project may be of use to interested parties.

Thank you!

Q&A Session