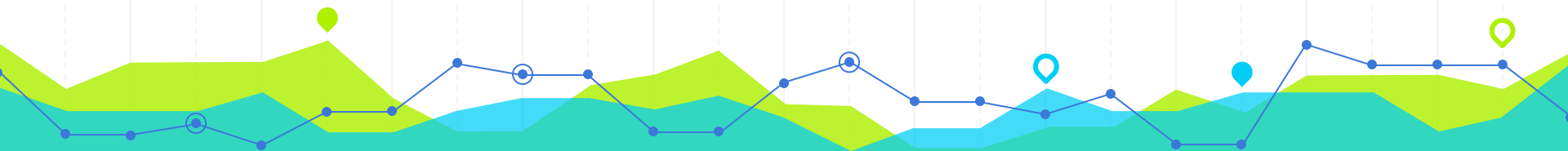


Exploring NYC Transit Data

Using open source tools



Midterm Project Presentation
DATASCI W209
March 10th, 2016

James | Drew | Greg

Introduction



The NYC Taxi & Limousine Commission (TLC) has made [publicly available](#) all green and yellow taxi trips between 2009 and 2015. This is a massive data source at > 1.1 Billion records!



Uber, by way of an information request submitted by 538 under FOIL, has [made available](#) NYC pickup data between April & September 2014 and January - July 2015, respectively .



U B E R

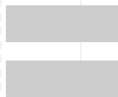


System Data is made [readily available](#) on Citi Bike Share trips

NYC OpenData

Relevant Open and accessible datasets include:

1. Weather
2. [Felony Incidents](#)
3. [MTA information - turnstile usage](#)



Visual Research Potential

A unique opportunity to:

- Build on some of the unique and amazing work done by:
 - [538](#),
 - [Chris Whong](#)
 - [Todd Schneider](#)
- Leverage techniques and frameworks taught in the course: downsampling, context, form and D3 respectively.
- Really tell a story through data - explore questions about NYC *in a way not traditionally possible*



Goal & Motivations

Acquire, integrate and analyze NYC open transit data to uncover macro transit trends that can be further explored in a narrow context to tell a story about a specific group of people

Hypothesis

- We hypothesize that robust well integrated transit datasets can answer real questions about how New Yorkers (*read: humans*) live, work and play
- Any increase or decrease in demand for NYC transit modalities has far reaching implications given transportation's importance and proxy for economic activity (eg. 175 million taxi rides in 2015)
- NYC is the perfect petri dish for analysis with its sky high population and transport density

Applicability

- We seek to visually explore both broad and narrow conceptual questions
- **Broadly**
 - Can we pinpoint the efficiency of the transit system at various points in time?
 - Can we visually infer how exogenous changes influence modality choice / driver behavior?
- **Narrowly**
 - Can we tell a story about NYU students and their economic behavior given transit choice and options?*

*to be further developed / evolved



Architecture / Process Overview

Data Sources



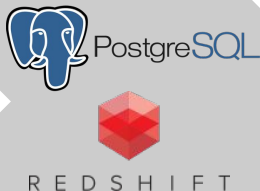
- Download entirety of NYC Taxi, Uber, Citibike and other selected open data via custom bash scripts
- Parse & clean, standardize information and load into a `Trips` schema based on dataset conformity and relevancy

Acquisition

`#!/bin/bash`



Storage



- Preprocess into PostgreSQL & PostGIS to obtain geometry data
- Join main transit data together and on attributes from other open data (Weather, Felony Incidents and MTA information)
- Load processed data into Redshift for high performance columnar analysis ability
- Leverage R / Python to investigate particular questions

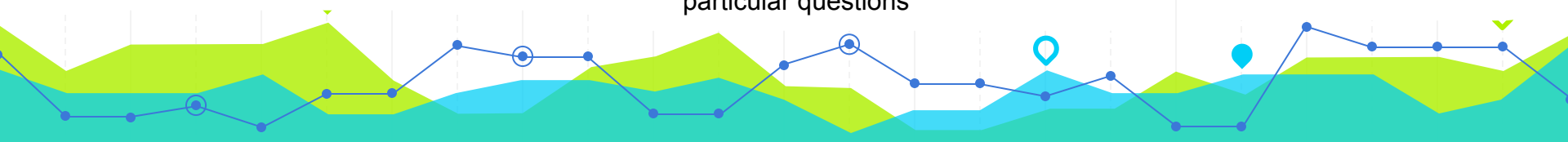
Computation



Visualization



- Create initial prototype in Tableau
- Visualize final results **as a web-app** via Bokeh or Shiny & leveraging D3 for map visuals



ER Diagram

This is the present model for the NYC transit team. It is hosted in Redshift.

Blue Outline: Main trip fact table

Green Outline: Dimension tables related to the fact table

Red Outline: Non-conforming Uber data from 2015, 2014 Uber data is in table Trips

Yellow Outline: Derived - record counts

Not yet shown:

1. Other open data (Felonies, MTA, etc)

2. Citibike Data

central_park_weather_observations_raw	
station_id	character varying N
station_name	character varying N
date	date
precipitation_tenths_of_mm	numeric N
snow_depth_mm	numeric N
snowfall_mm	numeric N
max_temperature_tenths_degrees_celsius	numeric N
min_temperature_tenths_degrees_celsius	numeric N
average_wind_speed_tenths_of_meters_per_second	numeric N

record_counts	
cnts	bigint N
mnth	double precision N
cab_type_integer	integer N

nyct2010	
gid	integer
ctlabel	character varying(7) N
borocode	character varying(1) N
boroname	character varying(32) N
ct2010	character varying(6) N
boroc2010	character varying(7) N
cdellgibil	character varying(1) N
ntacode	character varying(4) N
ntaname	character varying(75) N
puma	character varying(4) N
shape_leng	numeric
shape_area	numeric

trips	
id	integer
cab_type_id	integer N
vendor_id	character varying N
pickup_datetime	timestamp N FK
dropoff_datetime	timestamp N
store_and_fwd_flag	character(1) N
rate_code_id	integer N
pickup_longitude	numeric N
pickup_latitude	numeric N
dropoff_longitude	numeric N
dropoff_latitude	numeric N
passenger_count	integer N
trip_distance	numeric N
fare_amount	numeric N
extra	numeric N
mta_tax	numeric N
tip_amount	numeric N
tolls_amount	numeric N
ehail_fee	numeric N
improvement_surcharge	numeric N
total_amount	numeric N
payment_type	character varying N
trip_type	integer N
pickup_nyct2010_gid	integer N
dropoff_nyct2010_gid	integer N

uber_trips_2015	
id	integer
dispatching_base_num	character varying N
pickup_datetime	timestamp N
affiliated_base_num	character varying N
location_id	integer N
nyct2010_ntacode	character varying N

uber_taxi_zone_lookups	
location_id	integer
borough	character varying N
zone	character varying N
nyct2010_ntacode	character varying N

cab_types	
id	integer
type	character varying N

Target Audience

- Urban Planners
 - “Do trends in traffic flow predict a need for new / different zoning or better transportation infrastructure?”
- Traffic Engineers
 - “Does indicated traffic flow conflict with existing traffic constraints such as large impending construction projects?”
- Sociologists / policy-makers / economists
 - “Do trends in taxi vs. Uber / Lyft usage serve the public in the best way?”
- Data scientists / visualization enthusiasts
 - “What cool examples can I see for visualizing data?”



Data Sources

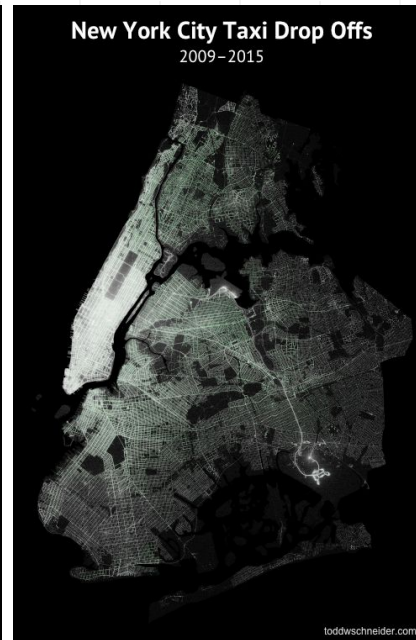


Synopsis

This is the granddaddy dataset and is extremely comprehensive in both size and breadth including: VendorId,

Summary Information

- URL: [Taxi TLC Data](#) / [DD](#)
- Date Range
 - Yellow: 2009 - 2015
 - Green: Aug 2013 - 2015
- Total Files
 - 116
- File Size (Total // Average):
 - 267G // 2.3G
- Average Record Count:
 - 14M



Prior Work done by and self published by Todd Schneider
[\[Image attribution\]](#)

Uber Data U B E R

Synopsis

This dataset is fairly limited and only includes: date and time of pickup, lat/long, and a base code that corresponds to a TLC station name

Summary Information

- URL: [Uber Data](#)
- Date Range
 - July 2014 - September 2014
 - Jan 2015 - June 2015
- Total Files
 - 7
- File Size (Total // Average):
 - 727M // 103M
- Average Record Count:
 - 1M

Manhattan Dominates Both Ubers And Cabs

Residential pickup rates by borough

BOROUGH	PICKUP INDEX (100 = AVG.)	
	UBER	CABS
Bronx	8	9
Brooklyn	79	29
Manhattan	357	431
Queens	19	22
Staten Island	1	0

Pickup data from April through Sept. 2014



SOURCE: TAXI & LIMOUSINE COMMISSION

Lower Income Means Fewer Pickups

Residential pickup rates by median income of census tract

MEDIAN INCOME	PICKUP INDEX (100 = AVG.)	
	UBER	CABS
\$0 - 25K	21	26
25 - 50	32	39
50 - 75	75	67
75 - 100	160	167
100 - 125	419	462
125 - 150	649	725
150+	539	564

Pickup data from April through Sept. 2014



TAXI & LIMOUSINE COMMISSION, CENSUS BUREAU

Some prior work done by and published on <http://fivethirtyeight.com/> [Image attribution]

Bikeshare Data



Synopsis

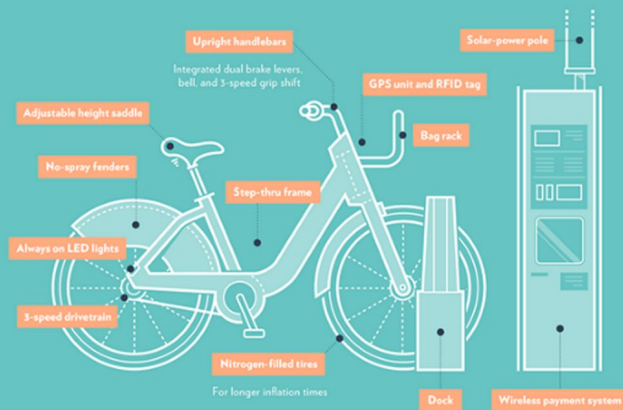
This dataset provides trip history information including: Duration, Start and End Time, Station Names & Lat / Long, Bike ID, User Type, Gender & Birthday

Summary Information

- URL: [System Data](#)
- Date Range
 - July 2013 - December 2015
- Total Files
 - 30
- File Size (Total // Average):
 - 4G // 130M
- Average Record Count:
 - 1M

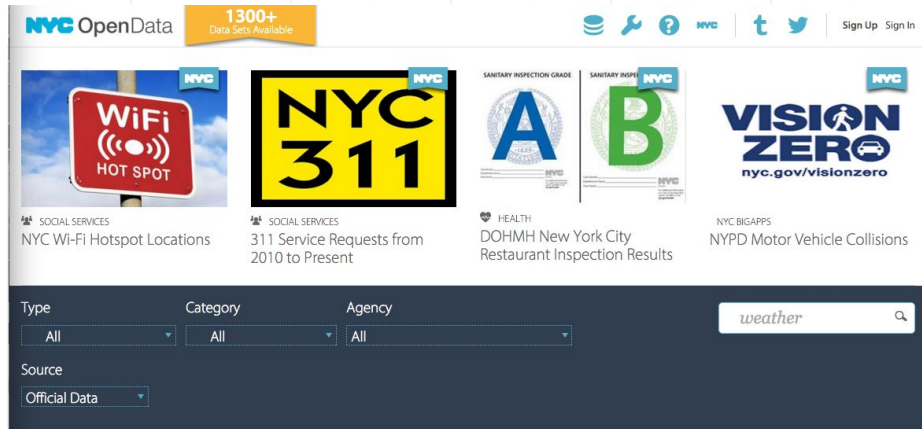
The Bike

What makes Citi Bike bicycles unique?



For those not familiar w/ bike share data, the units when docked roughly look like this. [\[Image attribution\]](#)

NYC Open Data



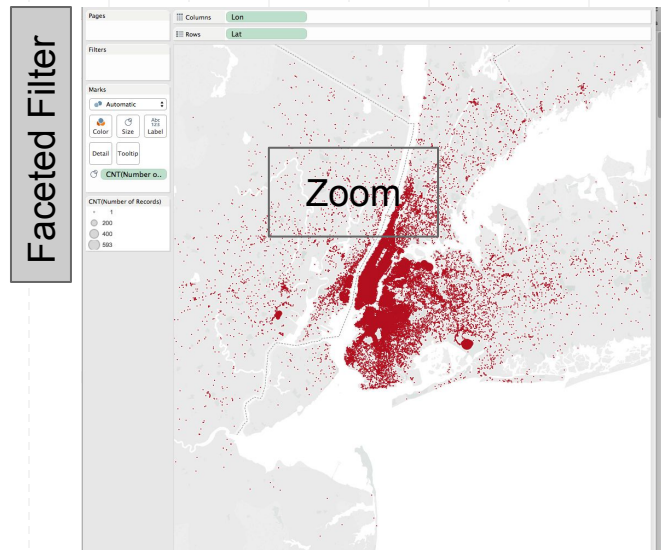
*to be further developed / evolved

- **Crime:** using long/lat location to evaluate impact on transit
- **Weather:** how does climate conditions impact transit
- **Subway/Bus/Rail:** volume, location implications on other forms of transit

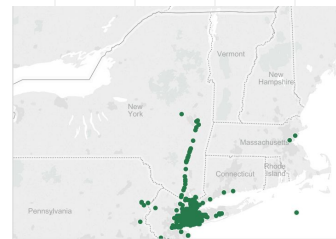
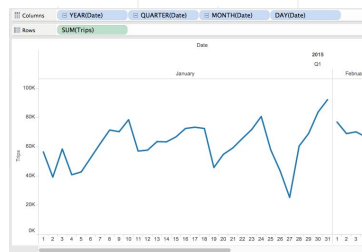
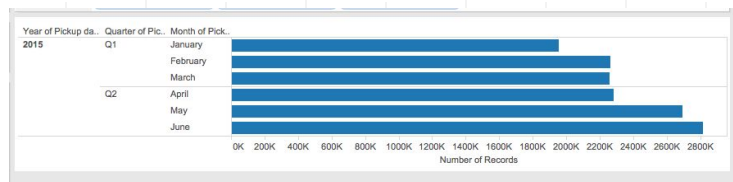
Current Mockup



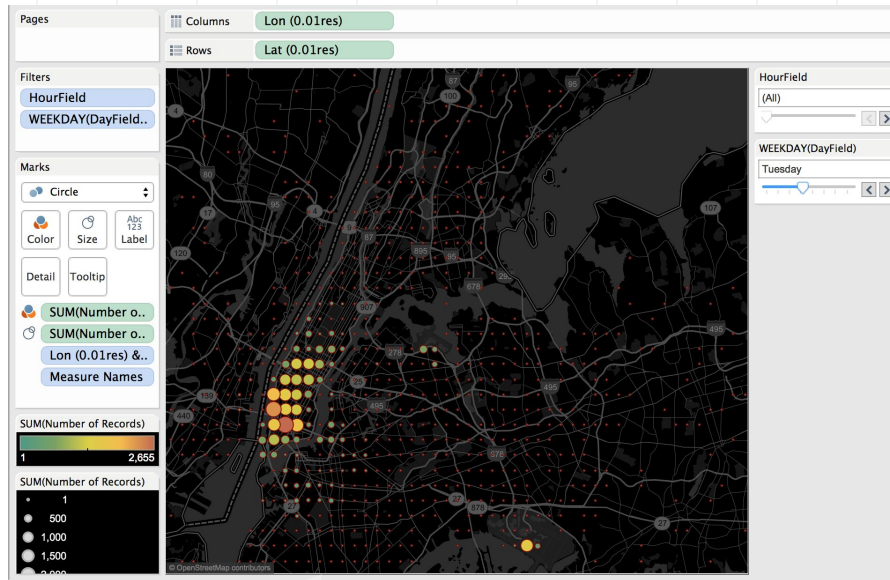
Anticipated Interactions



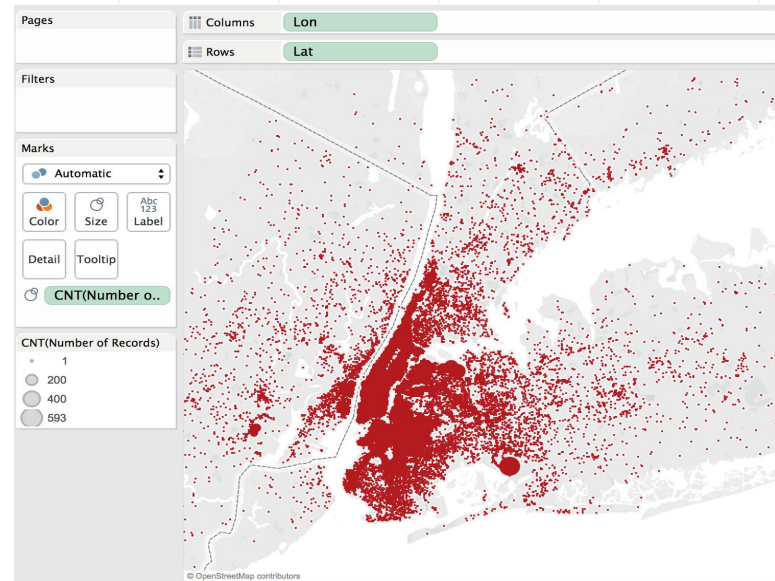
- Overview, zoom, filter, details-on-demand
- Navigate from broad to specific analyses
- Potential visualizations optimized for personas/tasks
- Interactivity: brushing and linking, overview + detail, zooming and panning



Mockup - Iteration #1



Some work done to explore and aggregate counts by lat/long pairs over time



An additional attempt to map and show how disbursed all Uber pickups are

Mockup - Iteration #2

NYC OpenData

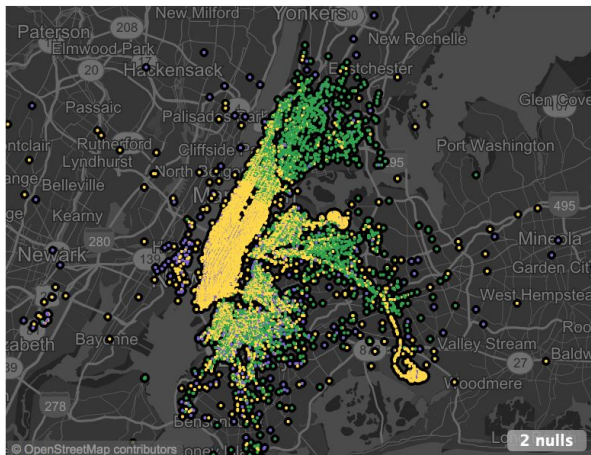
2014

April

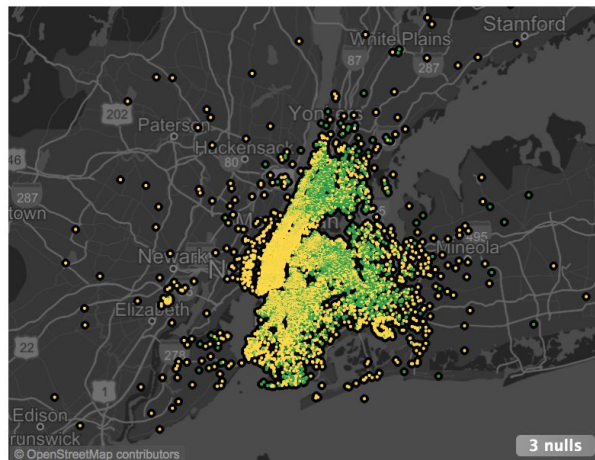
Company

(All)

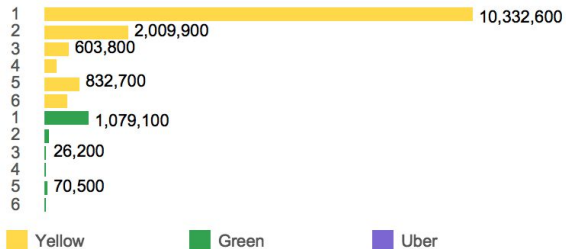
Pickups



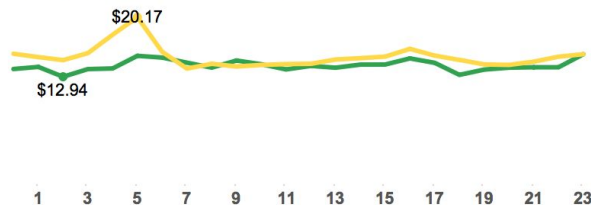
Dropoffs



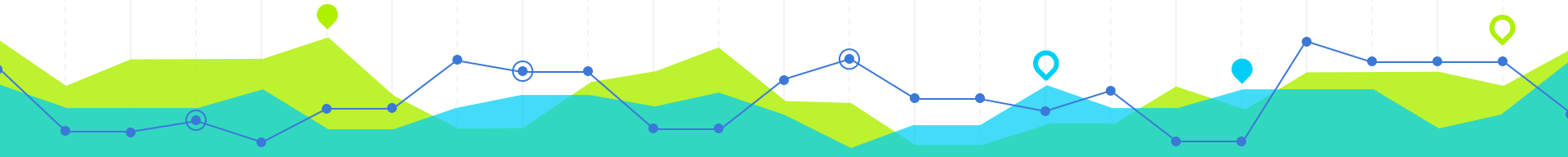
Trip Counts by Num of Passengers



Hourly Average Fare By Hour



Testing Plan



Expected Feedback Capture



Initial Thoughts (Verbatims):

1. Without clicking on anything, what do you expect each drop-down select to do?
2. Examine the titles for each of the visualizations. What do you think each title means?
3. What do you expect will happen if you choose a []? What about after choosing more than one?

Interactive Exercises

1. Depending on final visualization form, intend on capturing user feedback that requires interactive exploration and discovery
2. Likely include 5 - 10 "challenges" for the user to complete

Usability Questions

1. Test questions along the lines of: the following will be administered:
 - a. User Satisfaction
 - b. Usage Simplicity
 - c. Productivity
 - d. Useful Error Messaging
 - e. Ease of Navigation
 - f. Information Organization
 - g. Meeting expectations

Final Analyses & Feedback

1. Was the layout intuitive? If not, how would you adjust the layout?
2. Beyond the questions listed above, what improvements could be made to the tool?
3. What do you wish the tool could do for you?

Sending Us Feedback

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Questions?

