# **Project Premise**

CitiBank has a New York City bike service where many bike stations are available to the general public. Any user may rent out a bike for a time to ride around the city. There is also a subscription service for these bikes where users can pay a monthly fee for an unlimited use per year. Many different people use this bike service. Are there unifying variables, which connect all these users? What users is this service more popular with and how can a new market of people be convinced to use this bike service? How can CitiBank market using deals or ads?

# **Data Source:**

The data comes from a single dataset, Citi Bike Data by Kaggle user Ryan Cummings. The dataset can be viewed via Kaggle <a href="here">here</a>. The Spotify and Youtube dataset includes data from various NYC Citi Bike users. The data is from October 2013, but the Kaggle page shows the page was updated four years ago. The dataset contains statistics from 50,000 different trips.

# **Cleaning Data:**

Column	Issue	Action
trip_id	Unnecessary for analysis	Deleted
bike_id	Unnecessary for analysis	Deleted
weekday	Confusing column name	Changed name to 'day_of_week'
gender	Column values are not quick to interpret	Changed data type from int to str and then replaced the number values with the following: N/A, M, F

## **Understanding Data:**

Variable	Description	
trip_id	Bike trip's individual id	
bike_id	Bike's individual id	
weekday	Day of the week the bike trip occurred	
start_hour	Hour of the day the bike trip occurred	
start_time	Date and time the bike trip started	

start_station_id	Station id of the station where the bike trip started
start_station_name	Station name where the bike trip started
start_station_latitude	Station latitude where the bike trip started
start_station_longitude	Station longitude where the bike trip started
end_time	Date and time the bike trip ended
end_station_id	Station id of the station where the bike trip ended
end_station_name	Station name where the bike trip ended
end_station_latitude	Station latitude where the bike trip ended
end_station_longitude	Station longitude where the bike trip ended
trip_duration	Trip's duration measured in seconds
subscriber	User's status on whether or not they are a CitiBike subscriber
birth_year	User's birth year, non-subscribers do not enter their birth year
gender	User's gender (0 = unknown, 1 = male, 2 = female)

Variable	Time-Variant or Invariant	Qualitative or Quantitative	Binary or Nominal or Ordinal	Discrete or Continuous
trip_id	Invariant •	Qualitative •	Nominal -	N/A -
bike_id	Invariant •	Qualitative •	Nominal •	N/A ·
weekday	Invariant •	Qualitative •	Nominal •	N/A ·
start_hour	Time Variant •	Qualitative •	Nominal •	N/A ·
start_time	Time Variant -	Qualitative •	Nominal -	N/A -
start_station_id	Invariant •	Qualitative •	Nominal •	N/A ·
start_station_name	Invariant •	Qualitative •	Nominal -	N/A -
start_station_latitude	Invariant •	Qualitative •	Nominal •	N/A -

start_station_longitude	Invariant •	Qualitative -	Nominal •	N/A -
end_time	Time Variant -	Qualitative -	Nominal •	N/A -
end_station_id	Invariant •	Qualitative -	Nominal •	N/A +
end_station_name	Invariant •	Qualitative -	Nominal •	N/A +
end_station_latitude	Invariant •	Qualitative -	Nominal •	N/A -
end_station_longitude	Invariant •	Qualitative -	Nominal •	N/A -
trip_duration	Time Variant -	Quantative •	N/A ·	Continuous •
subscriber	Time Variant -	Qualitative •	Nominal •	N/A -
birth_year	Invariant •	Qualitative •	Ordinal •	N/A -
gender	Invariant •	Qualitative •	Binary -	N/A -

	bike_id	start_hour	start_station_id	$start\_station\_latitude$	start_station_longitude	end_station_id	end_station_latitude	end_station_longitude
count	50000.000000	50000.000000	50000.000000	50000.000000	50000.000000	50000.000000	50000.000000	50000.000000
mean	17615.269360	14.145240	443.321500	40.734170	-73.991109	442.539700	40.733859	-73.991351
std	1675.407446	4.860541	356.559925	0.019911	0.012555	355.756022	0.019885	0.012569
min	14556.000000	0.000000	72.000000	40.680342	-74.017134	72.000000	40.680342	-74.017134
25%	16188.000000	10.000000	304.000000	40.720196	-74.000271	304.000000	40.720196	-74.001547
50%	17584.000000	15.000000	402.000000	40.735877	-73.990765	402.000000	40.735354	-73.991218
75%	19014.000000	18.000000	484.000000	40.750020	-73.981923	483.000000	40.749013	-73.982050
max	20642.000000	23.000000	3002.000000	40.770513	-73.950048	3002.000000	40.770513	-73.950048

gender	birth_year	trip_duration
50000.000000	43021.000000	50000.000000
1.073540	1975.627786	838.982900
0.589389	11.089001	573.663997
0.000000	1899.000000	60.000000
1.000000	1968.000000	417.000000
1.000000	1978.000000	672.000000
1.000000	1984.000000	1112.000000
2.000000	1997.000000	2697.000000

## Limitations:

The dataset is current as of October 2013, so the data may not reflect current user trends. Variables such as user IDs and trip costs are not included. This excluded variables prevents any analysis of repeat customers, such as seeing how much they use the CitiBike service. Trip costs can be used to analyze how much cost factors into how users decide on how to ride on a CitiBike.

## Ethics:

Data does not contain any HIPPA-related information. Data has been collected from CitiBike and does not make any mention of private information.

# **Key Questions:**

- What variable matters the most when people decide to ride a Citi Bike?
- When is the busiest time for Citi Bike use?
- What is the busiest day for Citi Bike?
- What are the characteristics of an average Citi Bike user? (Age, gender, etc)
- How long do people usually ride for? Do different scenarios change ride times? (Day of the week, time, etc)
- Which stations are people leaving from and arriving at?
- How different are subscriber statistics vs non-subscriber statistics?