

# Increasing Attendance at the WTWY Summer Gala

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# Objectives

WomenTechWomenYes wants to optimize placement of street teams around subway stations, in order to:

- Maximize Signups
- Maximize Attendance
- Maximize Donations

# Target Audience Profile



← This is Becky.

- 20 to 30-something woman
- Works for a tech company
- Enjoys Starbucks coffee

# Approach

Two part approach:

1. Maximize total number of people seen by street teams

- Busiest\* subway stations (MTA Data)

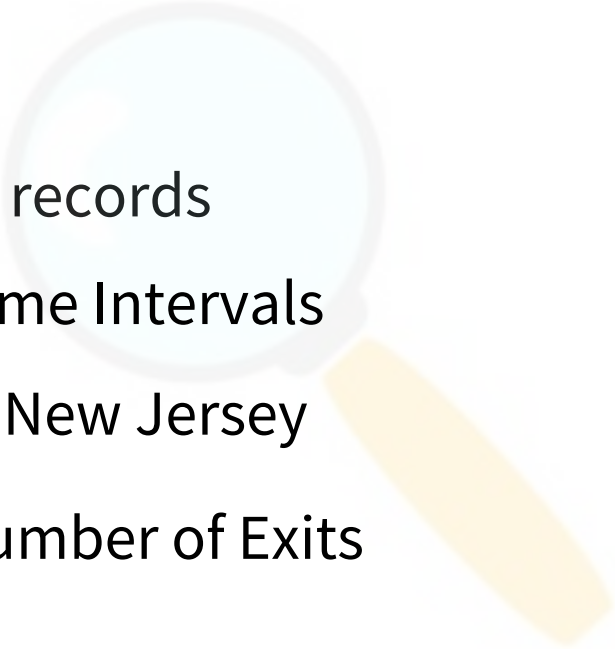
2. Maximize the likelihood of signups & attendance

- Station proximity to a Starbucks (Google Maps Data)
- Station proximity to Tech companies (Google Maps Data)
- Percentage gender score for each station (US Census Data)

\*Stations with the most exits

# MTA Data Exploration

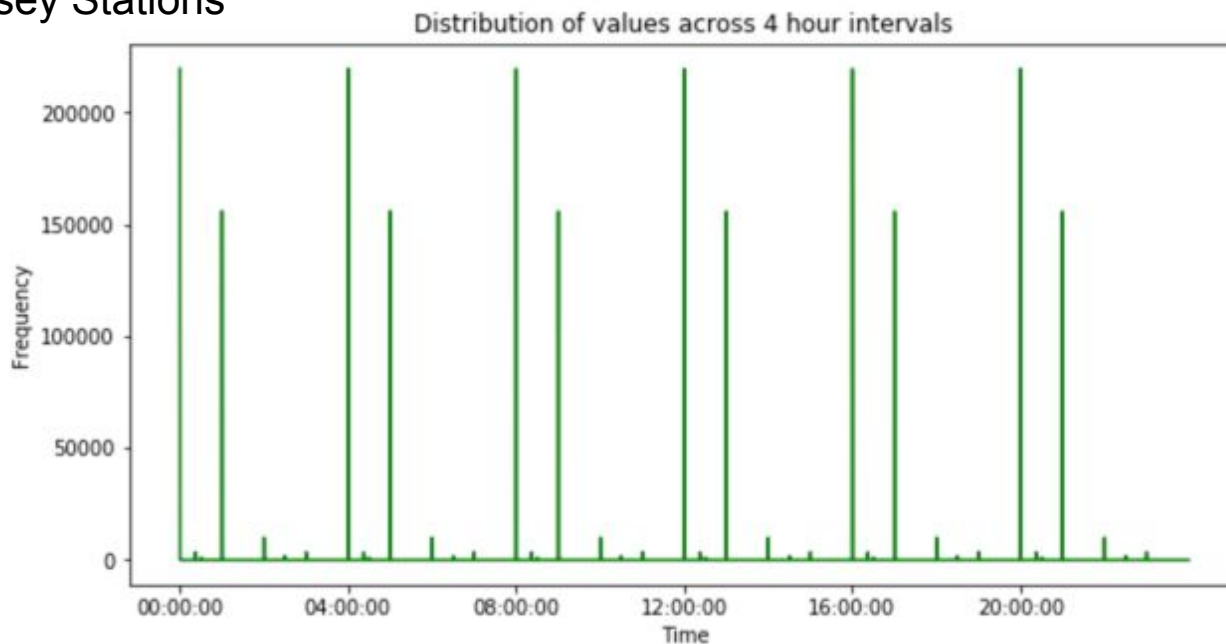
## We Identified:

- Duplicated records
  - Random Time Intervals
  - Stations in New Jersey
  - Extreme Number of Exits
- 

# MTA Data Cleaning

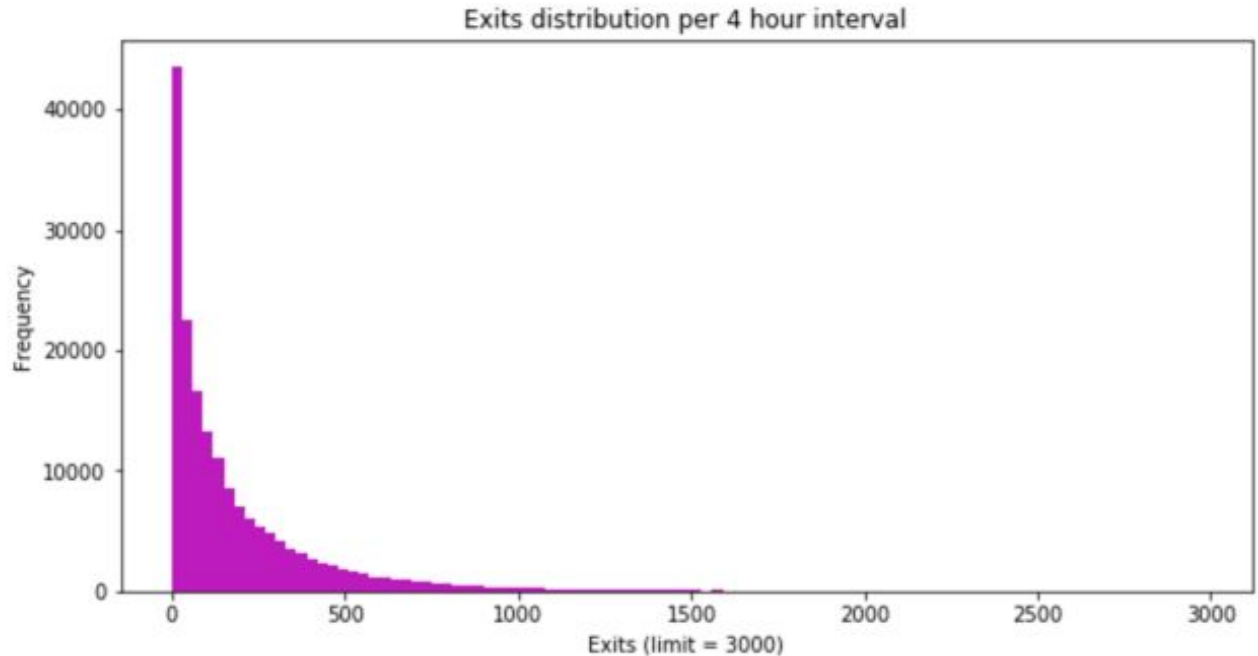
## Records at random minutes:seconds

- 75% New Jersey Stations



# MTA Data Cleaning

Extreme values:



# MTA Analysis - Interactive Interface

# LET'S SEE IT IN ACTION!

```
In [ ]: activity, dct, df = main(month, day, hour, yrs back)
```

```
In [282]: activity by time(5,9,dct)[:10]
```

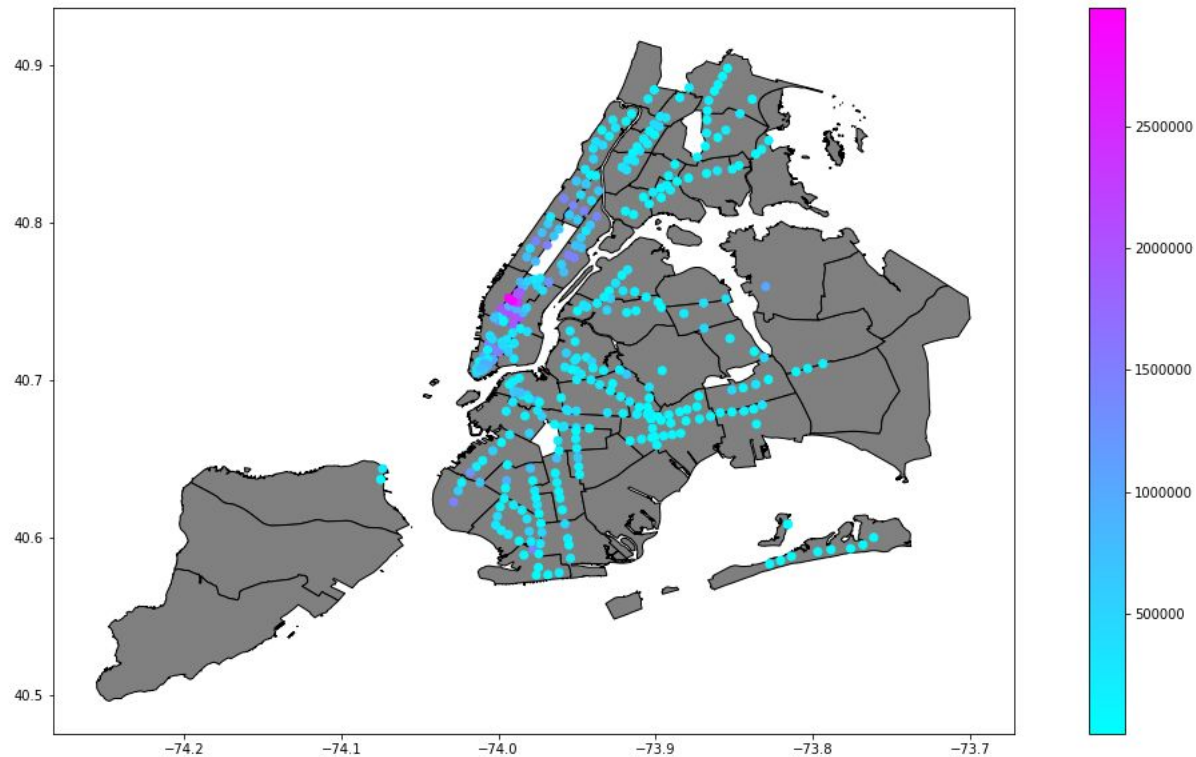
```
Out[282]: [['TIMES SQ-42 ST', 4257.166666666667],
['14 ST-UNION SQ', 3772.3333333333335],
['34 ST-HERALD SQ', 3469.1875],
['59 ST COLUMBUS', 1797.125],
['FLUSHING-MAIN', 1589.3125],
['W 4 ST-WASH SQ', 1574.25],
['59 ST', 1567.5625],
['8 AV', 1424.1875],
['BEDFORD AV', 1235.25],
['B'WAY-LAFAYETTE', 1229.9375]]
```

```
clean_df.cum(dct)
```

```
array([[ 'TIMES SQ-42 ST', 3151.29935515873],
       ['14 ST-UNION SQ', 2986.3298611111113],
       ['34 ST-HERALD SQ', 2375.5850694444443],
       ['FLUSHING-MAIN', 2036.685267857143],
       ['ATL AV-BARCLAY', 1687.7100694444446],
       ['JKSN HT-ROOSVLT', 1686.421875],
       ['59 ST COLUMBUS', 1648.171875],
       ['BEDFORD AV', 1446.298859126984],
       ['59 ST', 1341.6450892857142],
       ['W 4 ST-WASH SQ', 1333.3482142857142],
       ['50 ST', 1105.7150207619048],
```



# MTA - Heat Map

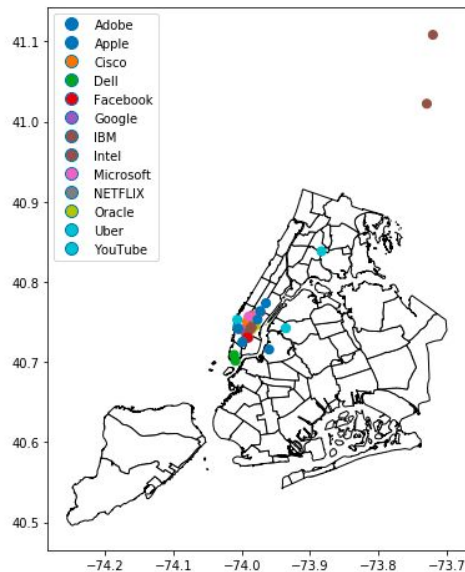


# Work - Tech Companies

- Top 21 most valuable American technology companies
- Use Google Places API to search for each company's location in New York City
- Calculate tech company proximity score for each station

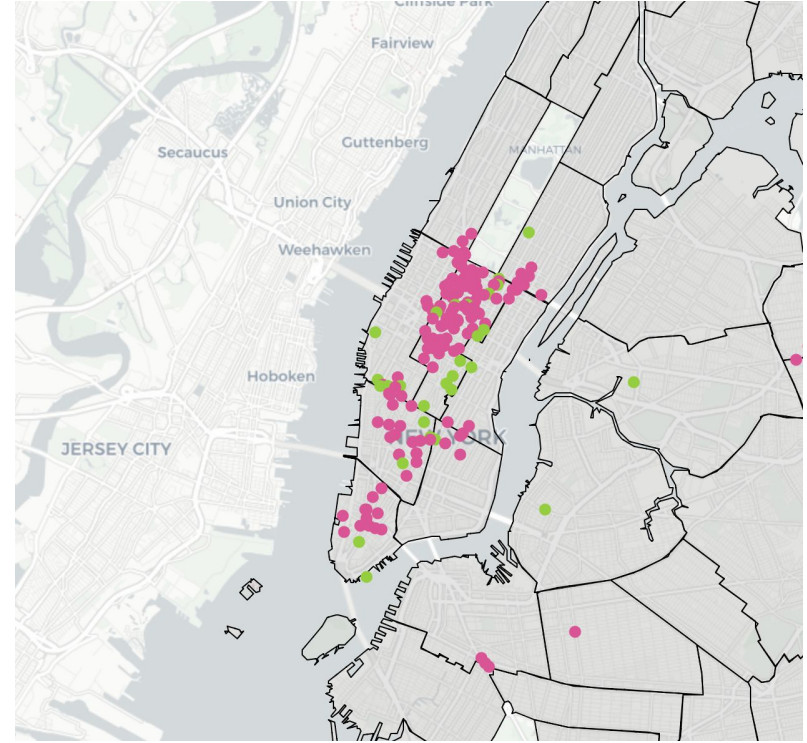
## The 21 most valuable tech brands in America

Myelle Lansat Jun 25, 2018, 11:44 AM



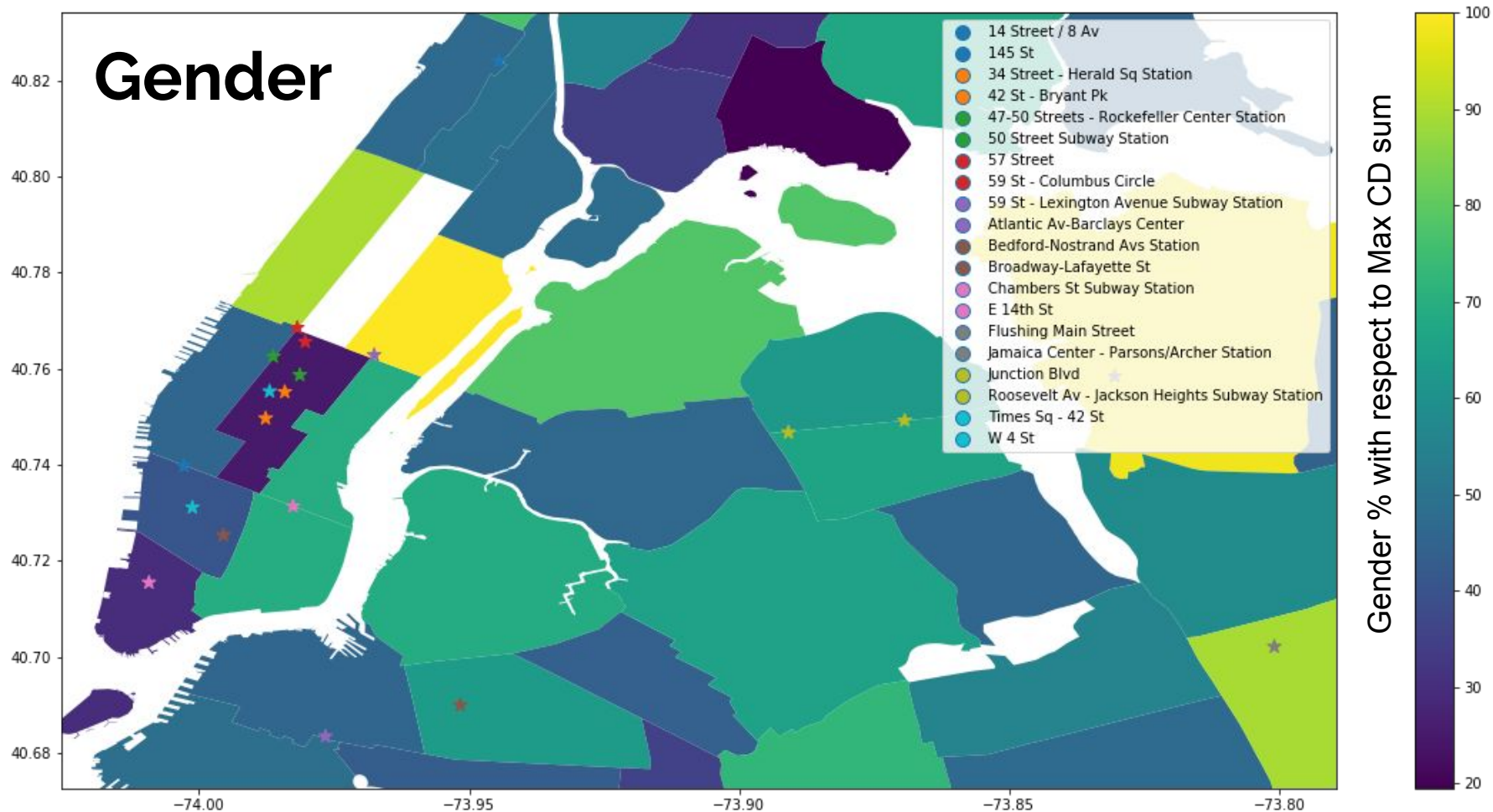
# Interests - Starbucks

- All Starbucks locations within 600 meters of the identified MTA stations were found
- Subway stations were scored based on the quantity and distance to Starbucks locations



- Green - tech companies
- Pink - Starbucks

# Gender



# Final Scoring

Equation used for scoring:

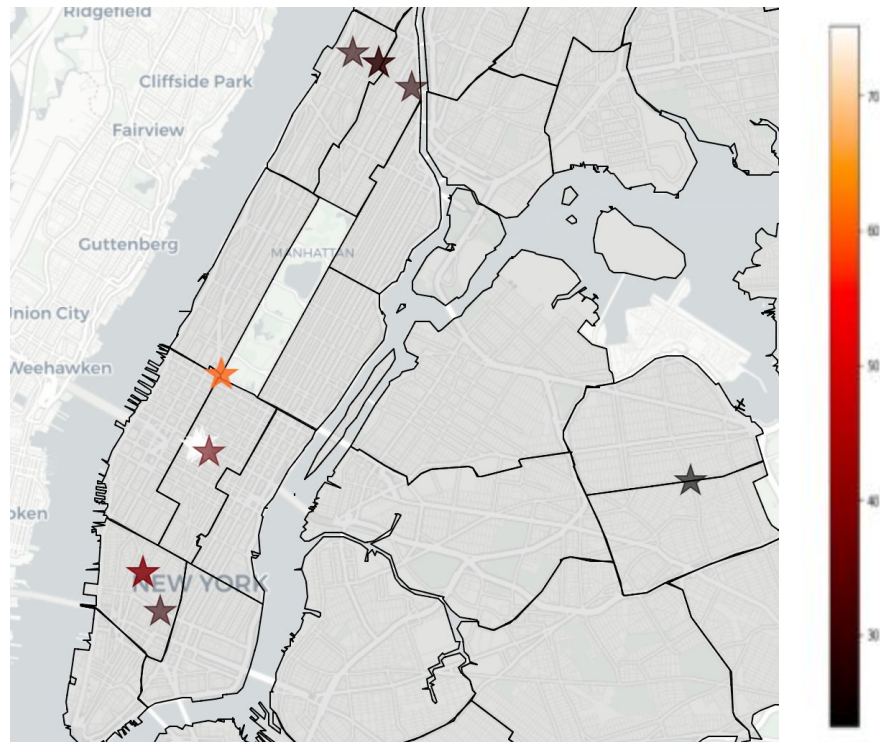
$$FS = W_1 * Score_{MTA} + W_2 * Score_{Tech} + W_3 * Score_{Coffee} + W_4 * Score_{Gender}$$

- $W_1 = 0.5$
- $W_2, W_3 = 0.15$
- $W_4 = 0.2$

# Overall Recommendations

Top 10 list of recommended stations:

1. Times Sq- 42 St
2. E 14th St
3. 34th St - Herald Sq Station
4. 59th St - Columbus Circle
5. Flushing Main Street
6. 59th St - Lexington Avenue Subway Station
7. 47 - 50 Rockefeller Center Station
8. 14th Street / 8 Av
9. W 4 St
10. Atlantic Av - Barclays Center



**Thank You!**

# Assumptions

- Timeline of analysis is now, summer time
- In the summer season, we expected that a majority of volunteers would be available between 12pm and 12am. We examined data in June for the past three years as a subset



# MTA Analysis: Final Product

- User inputs month, day, hour and number of years to scrape.
- Program provides three main functions:

# Takeaway Points

- MTA analysis → 'Busiest' stations mostly located south of Central Park: Midtown, Flatiron, SoHo, Financial District
- Starbucks and tech company proximity scores, as well as demographic data can be used to further prioritize busiest stations

# Next Steps

- Explore non-linear regression models
- Short term vs. long term engagement
- Continue to collect feedback and learning
- Other factors to consider
  - Tech awareness
  - Education level
  - Income level

# Results Table

	Subway_Station	MTA Score Percent (%)	Tech Company Percent (%)	Strbks Distance Score Percent (%)	Gender Percent (%)	Final_Score
0	Times Sq - 42 St	100.000000	55.670240	78.772617	25.018381	75.170105
1	E 14th St	94.857815	38.314692	40.158419	69.226205	73.045115
2	34 Street - Herald Sq Station	75.458110	54.950511	84.839765	25.018381	63.701273
3	59 St - Columbus Circle	52.320407	30.560302	77.566706	89.901583	60.359572
4	Flushing Main Street	64.848899	5.554432	27.907102	98.332194	57.110118
5	59 St - Lexington Avenue Subway Station	42.558717	37.314237	61.888235	100.000000	56.159729
6	50 Street Subway Station	35.128152	40.312335	93.198653	45.682149	46.727154
7	47-50 Streets - Rockefeller Center Station	18.413980	100.000000	94.633298	25.018381	43.405661
8	14 Street / 8 Av	23.446992	79.833552	67.421549	41.068273	42.025416
9	W 4 St	42.352487	40.626471	38.419252	41.068273	41.246757
10	Atlantic Av-Barclays Center	54.102154	10.969329	21.809874	45.754382	41.118834
11	57 Street	30.172809	36.908444	100.000000	25.018381	40.626347
12	42 St - Bryant Pk	20.141195	56.020324	95.494891	25.018381	37.801556
13	Roosevelt Av - Jackson Heights Subway Station	53.567523	8.978976	3.751323	44.085286	37.510364
14	Bedford-Nostrand Aves Station	45.940253	10.830579	0.000000	63.784230	37.351559
15	Jamaica Center - Parsons/Archer Station	33.664250	4.453061	7.093317	89.389503	36.441983
16	Chambers St Subway Station	22.933191	23.127369	83.942019	29.619358	33.450875
17	Broadway-Lafayette St	23.711866	40.053577	24.795028	41.068273	29.796878
18	145 St	33.592474	7.912415	9.170600	46.881732	28.735036
19	Junction Blvd	18.748353	7.346592	0.000000	63.784230	23.233011

# Introducing NYC's Top 100 tech companies



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