

# EDA - MTA Wi-Fi Network Working Plan

October 9, 2021

## 1 Overview:

The Metropolitan Transportation Authority (MTA) is the entity responsible over New York state subway system, recently they require to install Wi-Fi network in the subway stations to minimize the inconveniences that might results due to the wait time before train arrival by keeping the daily commuters well connected. The MTA contracted us as a consultancy group to advise on a working plan for this project.

```
[1]: #First we will upload the dataset in the ipython environment to start the MTA_  
    ↪data exploration  
    #Using pandas and sqlalchemy libraries  
  
import pandas as pd  
from flask_sqlalchemy import SQLAlchemy  
from sqlalchemy import create_engine
```

```
[2]: #Get the MTA data within the desired datetime range  
  
def get_data(week_nums):  
    url = "http://web.mta.info/developers/data/nyct/turnstile/turnstile_{}.txt"  
    dfs = []  
    for week_num in week_nums:  
        file_url = url.format(week_num)  
        dfs.append(pd.read_csv(file_url))  
    return pd.concat(dfs)
```

```
[3]: # The dataset used running data for the months of July, August and Septemeber_  
    ↪of 2021  
data_weeks = [210703,210710,210717,210724,210731,  
               210807,210814,210821,210828,  
               210904,210911,210918,210925,  
               211002]  
get_data(data_weeks)
```

```
[3]:
```

	C/A	UNIT	SCP	STATION	LINENAME	DIVISION	DATE	\
0	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	
1	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	
2	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	

3	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021
4	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021
...	...	...	...	...	...	...	...
210206	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021
210207	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021
210208	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021
210209	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021
210210	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021

	TIME	DESC	ENTRIES \
0	00:00:00	REGULAR	7592792
1	04:00:00	REGULAR	7592804
2	08:00:00	REGULAR	7592816
3	12:00:00	REGULAR	7592870
4	16:00:00	REGULAR	7592992
...	...	...	...
210206	05:00:00	REGULAR	5554
210207	09:00:00	REGULAR	5554
210208	13:00:00	REGULAR	5554
210209	17:00:00	REGULAR	5554
210210	21:00:00	REGULAR	5554

	EXITS
0	2595706
1	2595713
2	2595729
3	2595762
4	2595791
...	...
210206	649
210207	649
210208	649
210209	649
210210	650

[2934629 rows x 11 columns]

```
[4]: #Storing the dataset in pandas dataframe to be able to manipulate it
my_df = get_data(data_weeks)
engine = create_engine("sqlite:///MTA_db.db")
my_df.to_sql('MTA_table', engine, if_exists = 'replace', index=False)
```

```
[5]: tables = engine.table_names()
print(tables)
```

['MTA\_table']

/var/folders/4y/plz6nn617g3gccj119\_90bhc0000gn/T/ipykernel\_3542/1698560817.py:1:

SADeprecationWarning: The Engine.table\_names() method is deprecated and will be removed in a future release. Please refer to Inspector.get\_table\_names().  
(deprecated since: 1.4)

```
tables = engine.table_names()
```

```
[6]: my_df=pd.read_sql('select * from MTA_table', engine)
my_df.tail()
```

```
[6]:
```

	C/A	UNIT	SCP	STATION	LINENAME	DIVISION	DATE	\
2934624	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021	
2934625	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021	
2934626	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021	
2934627	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021	
2934628	TRAM2	R469	00-05-01	RIT-ROOSEVELT	R	RIT	10/01/2021	

	TIME	DESC	ENTRIES	\
2934624	05:00:00	REGULAR	5554	
2934625	09:00:00	REGULAR	5554	
2934626	13:00:00	REGULAR	5554	
2934627	17:00:00	REGULAR	5554	
2934628	21:00:00	REGULAR	5554	

	EXITS
2934624	649
2934625	649
2934626	649
2934627	649
2934628	650

```
[7]: #Removing the leading and trailing whitespaces using strip() function
my_df.columns.str.strip()
```

```
[7]: Index(['C/A', 'UNIT', 'SCP', 'STATION', 'LINENAME', 'DIVISION', 'DATE', 'TIME',
        'DESC', 'ENTRIES', 'EXITS'],
        dtype='object')
```

```
[8]: #The dataframe below represents a snippet of the data (the first 4 rows)
my_df.head()
```

```
[8]:
```

	C/A	UNIT	SCP	STATION	LINENAME	DIVISION	DATE	TIME	\
0	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	00:00:00	
1	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	04:00:00	
2	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	08:00:00	
3	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	12:00:00	
4	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	16:00:00	

	DESC	ENTRIES	\
0	REGULAR	7592792	

```

1 REGULAR 7592804
2 REGULAR 7592816
3 REGULAR 7592870
4 REGULAR 7592992

```

EXITS

```

0 2595706
1 2595713
2 2595729
3 2595762
4 2595791

```

- We can create a new column to combine both the date and time

```
[9]: my_df['DATETIME'] = pd.to_datetime(my_df.DATE + ' ' + my_df.TIME)
```

```
[10]: my_df.head()
```

```
[10]:
```

	C/A	UNIT	SCP	STATION	LINENAME	DIVISION	DATE	TIME	\
0	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	00:00:00	
1	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	04:00:00	
2	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	08:00:00	
3	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	12:00:00	
4	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	16:00:00	

```

DESC ENTRIES \
0 REGULAR 7592792
1 REGULAR 7592804
2 REGULAR 7592816
3 REGULAR 7592870
4 REGULAR 7592992

```

EXITS

```

0 2595706
1 2595713
2 2595729
3 2595762
4 2595791

```

DATETIME

```

0 2021-06-26 00:00:00
1 2021-06-26 04:00:00
2 2021-06-26 08:00:00
3 2021-06-26 12:00:00
4 2021-06-26 16:00:00

```

```
[11]: #Check for NA vlaues in the dataset
#The following two runs shows that there are no observations of type 'NA'
```

```
my_df.isna()
```

```
[11]:
```

	C/A	UNIT	SCP	STATION	LINENAME	DIVISION	DATE	TIME	\
0	False	False	False	False	False	False	False	False	
1	False	False	False	False	False	False	False	False	
2	False	False	False	False	False	False	False	False	
3	False	False	False	False	False	False	False	False	
4	False	False	False	False	False	False	False	False	
...	...	...	...	...	...	...	...	...	
2934624	False	False	False	False	False	False	False	False	
2934625	False	False	False	False	False	False	False	False	
2934626	False	False	False	False	False	False	False	False	
2934627	False	False	False	False	False	False	False	False	
2934628	False	False	False	False	False	False	False	False	

	DESC	ENTRIES	\
0	False	False	
1	False	False	
2	False	False	
3	False	False	
4	False	False	
...	...	...	
2934624	False	False	
2934625	False	False	
2934626	False	False	
2934627	False	False	
2934628	False	False	

	EXITS	\
0	False	
1	False	
2	False	
3	False	
4	False	
...	...	
2934624	False	
2934625	False	
2934626	False	
2934627	False	
2934628	False	

	DATETIME
0	False
1	False
2	False
3	False
4	False

```
[2934629 rows x 12 columns]
```

```
[12]: C/A
      UNIT
      SCP
      STATION
      LINENAME
      DIVISION
      DATE
      TIME
      DESC
      ENTRIES
      EXITS
      DATETIME
      dtype: int64
```

```

[13]:
0      A002  R051  02-00-00      59 ST  NQR456W      BMT  06/26/2021
1      A002  R051  02-00-00      59 ST  NQR456W      BMT  06/26/2021
2      A002  R051  02-00-00      59 ST  NQR456W      BMT  06/26/2021
3      A002  R051  02-00-00      59 ST  NQR456W      BMT  06/26/2021
4      A002  R051  02-00-00      59 ST  NQR456W      BMT  06/26/2021

...      ...      ...      ...      ...      ...      ...
2934624  TRAM2  R469  00-05-01  RIT-ROOSEVELT      R      RIT  10/01/2021
2934625  TRAM2  R469  00-05-01  RIT-ROOSEVELT      R      RIT  10/01/2021
2934626  TRAM2  R469  00-05-01  RIT-ROOSEVELT      R      RIT  10/01/2021
2934627  TRAM2  R469  00-05-01  RIT-ROOSEVELT      R      RIT  10/01/2021
2934628  TRAM2  R469  00-05-01  RIT-ROOSEVELT      R      RIT  10/01/2021

```

1	04:00:00	REGULAR	7592804
2	08:00:00	REGULAR	7592816
3	12:00:00	REGULAR	7592870
4	16:00:00	REGULAR	7592992
...	...	...	...
2934624	05:00:00	REGULAR	5554
2934625	09:00:00	REGULAR	5554
2934626	13:00:00	REGULAR	5554
2934627	17:00:00	REGULAR	5554
2934628	21:00:00	REGULAR	5554

#### EXITS

0	2595706
1	2595713
2	2595729
3	2595762
4	2595791
...	...
2934624	649
2934625	649
2934626	649
2934627	649
2934628	650

#### DATETIME

0	2021-06-26 00:00:00
1	2021-06-26 04:00:00
2	2021-06-26 08:00:00
3	2021-06-26 12:00:00
4	2021-06-26 16:00:00
...	...
2934624	2021-10-01 05:00:00
2934625	2021-10-01 09:00:00
2934626	2021-10-01 13:00:00
2934627	2021-10-01 17:00:00
2934628	2021-10-01 21:00:00

[2934629 rows x 12 columns]

```
[14]: my_df.duplicated().sum()
```

```
[14]: 0
```

```
[15]: #Locating the duplicate rows if any
my_df.loc[my_df.duplicated(), :]
```

```
[15]: Empty DataFrame
      Columns: [C/A, UNIT, SCP, STATION, LINENAME, DIVISION, DATE, TIME, DESC,
      ENTRIES, EXITS
      DATETIME]
      Index: []
```

```
[16]: #Get the count of entries for each Turnstile and check for duplicates
      #The results shows that there are two entries for the same turnstile in the
      ↳ same DATETIME series
      my_df.groupby(["C/A", "UNIT", "SCP", "STATION", "DATETIME"]).ENTRIES.count().
      ↳ reset_index().sort_values("ENTRIES", ascending=False)
```

```
[16]:
```

	C/A	UNIT	SCP	STATION	DATETIME	ENTRIES
2911468	S101	R070	00-00-02	ST. GEORGE	2021-09-15 08:00:00	2
2911391	S101	R070	00-00-02	ST. GEORGE	2021-09-02 20:00:00	2
2911401	S101	R070	00-00-02	ST. GEORGE	2021-09-04 12:00:00	2
2911400	S101	R070	00-00-02	ST. GEORGE	2021-09-04 08:00:00	2
2911398	S101	R070	00-00-02	ST. GEORGE	2021-09-04 00:00:00	2
...	...	...	...	...	...	...
978170	N128	R200	00-00-03	EUCLID AV	2021-07-31 04:00:00	1
978171	N128	R200	00-00-03	EUCLID AV	2021-07-31 08:00:00	1
978172	N128	R200	00-00-03	EUCLID AV	2021-07-31 12:00:00	1
978173	N128	R200	00-00-03	EUCLID AV	2021-07-31 16:00:00	1
2934481	TRAM2	R469	00-05-01	RIT-ROOSEVELT	2021-10-01 21:00:00	1

[2934482 rows x 6 columns]

```
[17]: import datetime

      test = ((my_df['C/A'] == 'R516') &
      (my_df['UNIT'] == 'R291') &
      (my_df['SCP'] == '00-00-02') &
      (my_df['STATION'] == '33 ST-RAWSON ST') &
      (my_df['DATETIME'].dt.date == datetime.datetime(2021, 8 , 25).date()))
```

```
[18]: my_df[test].tail()
```

```
[18]:
```

	C/A	UNIT	SCP	STATION	LINENAME	DIVISION	DATE	\
1864708	R516	R291	00-00-02	33 ST-RAWSON ST	7	IRT	08/25/2021	
1864709	R516	R291	00-00-02	33 ST-RAWSON ST	7	IRT	08/25/2021	
1864710	R516	R291	00-00-02	33 ST-RAWSON ST	7	IRT	08/25/2021	
1864711	R516	R291	00-00-02	33 ST-RAWSON ST	7	IRT	08/25/2021	
1864712	R516	R291	00-00-02	33 ST-RAWSON ST	7	IRT	08/25/2021	

	TIME	DESC	ENTRIES	\
1864708	12:00:00	RECOVR AUD	1535620	
1864709	16:00:00	REGULAR	10991727	



1864710	16:00:00	RECOVR AUD	1535734
1864711	20:00:00	REGULAR	10991870
1864712	20:00:00	RECOVR AUD	1535925

#### EXITS

1864708	1649485
1864709	8495220
1864710	1649537
1864711	8495226
1864712	1649602

#### DATETIME

1864708	2021-08-25 12:00:00
1864709	2021-08-25 16:00:00
1864710	2021-08-25 16:00:00
1864711	2021-08-25 20:00:00
1864712	2021-08-25 20:00:00

- The results above shows that there are more than 1 row for the same turnstile per entries, the difference between the two entries are the DESC column which has a Regular audit and a Recovered Audit.
- For the sake of simplicity, the Recovered Audit columns will be dropped.

```
[19]: my_df.drop(my_df[my_df['DESC'] == 'RECOVR AUD'].index, axis=0, inplace=True)
```

```
[20]: #checking the drop was done correctly
my_df[test].tail()
```

```
/var/folders/4y/plz6nn617g3gccj1l9_90bhc0000gn/T/ipykernel_3542/2925128706.py:2:
```

```
UserWarning: Boolean Series key will be reindexed to match DataFrame index.
```

```
my_df[test].tail()
```

```
[20]:
```

	C/A	UNIT	SCP	STATION	LINE	NAME	DIVISION	DATE
1864703	R516	R291	00-00-02	33 ST-RAWSON	ST	7	IRT	08/25/2021
1864705	R516	R291	00-00-02	33 ST-RAWSON	ST	7	IRT	08/25/2021
1864707	R516	R291	00-00-02	33 ST-RAWSON	ST	7	IRT	08/25/2021
1864709	R516	R291	00-00-02	33 ST-RAWSON	ST	7	IRT	08/25/2021
1864711	R516	R291	00-00-02	33 ST-RAWSON	ST	7	IRT	08/25/2021

	TIME	DESC	ENTRIES
1864703	04:00:00	REGULAR	10991599
1864705	08:00:00	REGULAR	10991607
1864707	12:00:00	REGULAR	10991639
1864709	16:00:00	REGULAR	10991727
1864711	20:00:00	REGULAR	10991870

#### EXITS

1864703	8495017
---------	---------

1864705	8495122
1864707	8495203
1864709	8495220
1864711	8495226

	DATETIME
1864703	2021-08-25 04:00:00
1864705	2021-08-25 08:00:00
1864707	2021-08-25 12:00:00
1864709	2021-08-25 16:00:00
1864711	2021-08-25 20:00:00

```
[21]: #checking for duplicate observations again
my_df.groupby(["C/A", "UNIT", "SCP", "STATION", "DATETIME"]).ENTRIES.count().
      ↪reset_index().sort_values("ENTRIES", ascending=False)
```

```
[21]:
```

	C/A	UNIT	SCP	STATION	DATETIME	ENTRIES
0	A002	R051	02-00-00	59 ST	2021-06-26 00:00:00	1
1946622	R138	R293	00-06-00	34 ST-PENN STA	2021-09-17 14:00:00	1
1946602	R138	R293	00-06-00	34 ST-PENN STA	2021-09-14 06:00:00	1
1946603	R138	R293	00-06-00	34 ST-PENN STA	2021-09-14 10:00:00	1
1946604	R138	R293	00-06-00	34 ST-PENN STA	2021-09-14 14:00:00	1
...	...	...	...	...	...	...
973306	N128	R200	00-00-03	EUCLID AV	2021-09-13 00:00:00	1
973307	N128	R200	00-00-03	EUCLID AV	2021-09-13 04:00:00	1
973308	N128	R200	00-00-03	EUCLID AV	2021-09-13 08:00:00	1
973309	N128	R200	00-00-03	EUCLID AV	2021-09-13 12:00:00	1
2919911	TRAM2	R469	00-05-01	RIT-ROOSEVELT	2021-10-01 21:00:00	1

[2919912 rows x 6 columns]

- The following steps will focus on understanding the data and discover any anomalies

```
[22]: my_df.DESC.value_counts()
```

```
[22]: REGULAR    2919912
Name: DESC, dtype: int64
```

```
[23]: my_df.shape
```

```
[23]: (2919912, 12)
```

```
[24]: my_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2919912 entries, 0 to 2934628
Data columns (total 12 columns):
#   Column                                Dtype
---  -
---
```

```

0    C/A
object
1    UNIT
object
2    SCP
object
3    STATION
object
4    LINENAME
object
5    DIVISION
object
6    DATE
object
7    TIME
object
8    DESC
object
9    ENTRIES                                int64
10   EXITS                                int64
11   DATETIME
datetime64[ns]
dtypes: datetime64[ns](1), int64(2), object(9)
memory usage: 289.6+ MB

```

```

[25]: #looking at the descriptive values for the commulative Entries and Exits as a
      ↪ballpark number doesn't make sense,
      #so we need to get the unique daily Entries to understand the problem statment
      ↪better. P.S. the Exits won't be
      #looked at in this analysis as our focus is to get a sense of the traffic per
      ↪station per day and it can be deduced
      #fairly from the daily Entries only.
      my_df.describe()

```

```

[25]:          ENTRIES  \
count    2.919912e+06
mean     4.149555e+07
std      2.181855e+08
min      0.000000e+00
25%      2.203418e+05
50%      1.397294e+06
75%      5.997214e+06
max      2.147407e+09

          EXITS
count    2.919912e+06
mean     3.322832e+07

```

std	1.918885e+08
min	0.000000e+00
25%	1.033300e+05
50%	8.541220e+05
75%	3.974544e+06
max	2.133797e+09

- The following few steps is to create a new column for unique daily entries and to check for any anomalies

```
[28]: #creating a turnstile_id for simplicity of coding
my_df['Turnstile_ID'] = my_df['C/A'].astype(str) + '_' + my_df.UNIT.astype(str)
    ↪+ '_' + \
my_df.SCP.astype(str) + '_' + my_df.STATION.astype(str)
```

```
[30]: my_df.head()
```

```
[30]:
```

	C/A	UNIT	SCP	STATION	LINENAME	DIVISION	DATE	TIME	\
0	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	00:00:00	
1	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	04:00:00	
2	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	08:00:00	
3	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	12:00:00	
4	A002	R051	02-00-00	59 ST	NQR456W	BMT	06/26/2021	16:00:00	

	DESC	ENTRIES	\
0	REGULAR	7592792	
1	REGULAR	7592804	
2	REGULAR	7592816	
3	REGULAR	7592870	
4	REGULAR	7592992	

	EXITS	\
0	2595706	
1	2595713	
2	2595729	
3	2595762	
4	2595791	

	DATETIME	Turnstile_ID
0	2021-06-26 00:00:00	A002_R051_02-00-00_59 ST
1	2021-06-26 04:00:00	A002_R051_02-00-00_59 ST
2	2021-06-26 08:00:00	A002_R051_02-00-00_59 ST
3	2021-06-26 12:00:00	A002_R051_02-00-00_59 ST
4	2021-06-26 16:00:00	A002_R051_02-00-00_59 ST

```
[31]: #For the purpose of this analysis, looking at the traffic per station is enough
    ↪to take decisions where to start
```

```
#the work to install the Wi-Fi Network as opposed to looking at it at a granular
↳ level per Turnstile
```

```
station_df = my_df.groupby(['STATION', 'DATE'])['ENTRIES'].max().reset_index()
```

```
station_df
```

```
[31]:
```

	STATION	DATE	ENTRIES
0	1 AV	06/26/2021	370891078
1	1 AV	06/27/2021	370891152
2	1 AV	06/28/2021	370891228
3	1 AV	06/29/2021	370891317
4	1 AV	06/30/2021	370891419
...	...	...	...
37087	ZEREGA AV	09/27/2021	1308054
37088	ZEREGA AV	09/28/2021	1308726
37089	ZEREGA AV	09/29/2021	1309387
37090	ZEREGA AV	09/30/2021	1310068
37091	ZEREGA AV	10/01/2021	1310763

```
[37092 rows x 3 columns]
```

```
[37]: station_df.sort_values(['STATION', 'DATE'], ascending=True)
```

```
station_df
```

```
[37]:
```

	STATION	DATE	ENTRIES
0	1 AV	06/26/2021	370891078
1	1 AV	06/27/2021	370891152
2	1 AV	06/28/2021	370891228
3	1 AV	06/29/2021	370891317
4	1 AV	06/30/2021	370891419
...	...	...	...
37087	ZEREGA AV	09/27/2021	1308054
37088	ZEREGA AV	09/28/2021	1308726
37089	ZEREGA AV	09/29/2021	1309387
37090	ZEREGA AV	09/30/2021	1310068
37091	ZEREGA AV	10/01/2021	1310763

```
[37092 rows x 3 columns]
```

```
[38]: station_df[['PREV_DATE', 'PREV_ENTRIES']] = station_df.
↳ groupby(['STATION'])['DATE', 'ENTRIES'] \
.apply(lambda x: x.shift(periods = 1, axis = 0, fill_value = 0))
```

```
station_df
```

```
/var/folders/4y/plz6nn617g3gccj119_90bhc0000gn/T/ipykernel_3542/518037491.py:1:
FutureWarning: Indexing with multiple keys (implicitly converted to a tuple of
```

keys) will be deprecated, use a list instead.

```
station_df[['PREV_DATE', 'PREV_ENTRIES']] =
station_df.groupby(['STATION'])['DATE', 'ENTRIES'] \
```

```
[38]:
```

	STATION	DATE	ENTRIES	PREV_DATE	PREV_ENTRIES
0	1 AV	06/26/2021	370891078	0	0
1	1 AV	06/27/2021	370891152	06/26/2021	370891078
2	1 AV	06/28/2021	370891228	06/27/2021	370891152
3	1 AV	06/29/2021	370891317	06/28/2021	370891228
4	1 AV	06/30/2021	370891419	06/29/2021	370891317
...	...	...	...	...	...
37087	ZEREGA AV	09/27/2021	1308054	09/26/2021	1307428
37088	ZEREGA AV	09/28/2021	1308726	09/27/2021	1308054
37089	ZEREGA AV	09/29/2021	1309387	09/28/2021	1308726
37090	ZEREGA AV	09/30/2021	1310068	09/29/2021	1309387
37091	ZEREGA AV	10/01/2021	1310763	09/30/2021	1310068

[37092 rows x 5 columns]

```
[39]: #Now let add one more column in the dataframe to get the unique daily entries
station_df['DAILY_ENTRIES'] = station_df.ENTRIES - station_df.PREV_ENTRIES

station_df
```

```
[39]:
```

	STATION	DATE	ENTRIES	PREV_DATE	PREV_ENTRIES	\
0	1 AV	06/26/2021	370891078	0	0	
1	1 AV	06/27/2021	370891152	06/26/2021	370891078	
2	1 AV	06/28/2021	370891228	06/27/2021	370891152	
3	1 AV	06/29/2021	370891317	06/28/2021	370891228	
4	1 AV	06/30/2021	370891419	06/29/2021	370891317	
...	...	...	...	...	...	
37087	ZEREGA AV	09/27/2021	1308054	09/26/2021	1307428	
37088	ZEREGA AV	09/28/2021	1308726	09/27/2021	1308054	
37089	ZEREGA AV	09/29/2021	1309387	09/28/2021	1308726	
37090	ZEREGA AV	09/30/2021	1310068	09/29/2021	1309387	
37091	ZEREGA AV	10/01/2021	1310763	09/30/2021	1310068	

	DAILY_ENTRIES
0	370891078
1	74
2	76
3	89
4	102
...	...
37087	626
37088	672
37089	661
37090	681

37091                      695

[37092 rows x 6 columns]

```
[40]: #Now let check the descriptive figures for the unique daily entries
station_df['DAILY_ENTRIES'].describe()
```

```
[40]: count      3.709200e+04
      mean      3.526291e+06
      std      7.681926e+07
      min     -1.942719e+09
      25%      1.220000e+02
      50%      3.820000e+02
      75%      6.820000e+02
      max      2.147407e+09
      Name: DAILY_ENTRIES, dtype: float64
```

- The unique daily entries shows negative minimum value, which is not correct given that the daily entries should be cumulative

```
[41]: station_df.dtypes
```

```
[41]: STATION      object
      DATE        object
      ENTRIES     int64
      PREV_DATE   object
      PREV_ENTRIES int64
      DAILY_ENTRIES int64
      dtype: object
```

```
[42]: station_df.PREV_ENTRIES = station_df.PREV_ENTRIES.astype(int)
      station_df.DAILY_ENTRIES = station_df.DAILY_ENTRIES.astype(int)
```

```
[44]: #lets take a subset of the dataframe to check how are the rows of negative
      ↪ value looking like
```

```
station_df[station_df.DAILY_ENTRIES < 0]
```

```
[44]:
```

	STATION	DATE	ENTRIES	PREV_DATE	PREV_ENTRIES	\
295	104 ST	06/27/2021	1681050247	06/26/2021	1681050263	
296	104 ST	06/28/2021	1681050222	06/27/2021	1681050247	
297	104 ST	06/29/2021	1681050128	06/28/2021	1681050222	
298	104 ST	06/30/2021	1681050016	06/29/2021	1681050128	
299	104 ST	07/01/2021	1681049915	06/30/2021	1681050016	
...	...	...	...	...	...	
34644	TIMES SQ-42 ST	10/01/2021	1891091412	09/30/2021	1891091983	
35080	UTICA AV	08/10/2021	17025492	08/09/2021	102825308	
35117	UTICA AV	09/16/2021	17089093	09/15/2021	102825371	

36462	WINTHROP ST	08/20/2021	1961869	08/19/2021	19163917
36466	WINTHROP ST	08/24/2021	7185669	08/23/2021	19167678

```

DAILY_ENTRIES
295          -16
296          -25
297          -94
298         -112
299         -101
...
34644         -571
35080       -85799816
35117       -85736278
36462       -17202048
36466       -11982009

```

[2762 rows x 6 columns]

- I am observing two abnormal patterns here:
  1. The first one is the negative daily entries (which is most likely to be erroneous entries where passengers reversed the turnstile wheel or something of that sort).
  2. The second one is the unexpected big difference between the ENTRIES and the PREV\_ENTRIES, where ENTRIES is a lot more smaller than the PREV\_ENTRIES and here I will assume if something like this happens then the commulative counter has been reset.
- In the next few steps, we will be working on fixing these anomalies.

```
[54]: station_df.dtypes
```

```

[54]: STATION          object
      DATE            object
      ENTRIES         int64
      PREV_DATE       object
      PREV_ENTRIES    int64
      DAILY_ENTRIES   int64
      dtype: object

```

```
[57]: station_df['DATE'] = pd.to_datetime(station_df['DATE'])
```

```

#df['Date'] = pd.to_datetime(df['Date'])
#df['Month'] = df['Date'].dt.month

```

```
[58]: station_df['PREV_DATE'] = pd.to_datetime(station_df['PREV_DATE'])
```

```
[59]: station_df.dtypes
```

```

[59]: STATION          object
      DATE            datetime64[ns]

```



```

ENTRIES                int64
PREV_DATE              datetime64[ns]
PREV_ENTRIES           int64
DAILY_ENTRIES          int64
dtype: object

```

```

[66]: station_df[(station_df['STATION'] == 'UTICA AV') & (station_df['DATE'] >=
↳ '2021-08-01') \
      & (station_df['DATE'] <= '2021-08-31')]

#created a subset of the UTICA AV station for the month of August, where I saw
↳ one of the anomalies above;
#to try and understand the problem at hand

```

```

[66]:
      STATION  DATE  ENTRIES  PREV_DATE  PREV_ENTRIES  DAILY_ENTRIES
35071  UTICA AV  2021-08-01  102822537  2021-07-31      102822333           204
35072  UTICA AV  2021-08-02  102822960  2021-08-01      102822537           423
35073  UTICA AV  2021-08-03  102823496  2021-08-02      102822960           536
35074  UTICA AV  2021-08-04  102823961  2021-08-03      102823496           465
35075  UTICA AV  2021-08-05  102824368  2021-08-04      102823961           407
35076  UTICA AV  2021-08-06  102824770  2021-08-05      102824368           402
35077  UTICA AV  2021-08-07  102824993  2021-08-06      102824770           223
35078  UTICA AV  2021-08-08  102825159  2021-08-07      102824993           166
35079  UTICA AV  2021-08-09  102825308  2021-08-08      102825159           149
35080  UTICA AV  2021-08-10  17025492  2021-08-09      102825308      -85799816
35081  UTICA AV  2021-08-11  17027428  2021-08-10      17025492           1936
35082  UTICA AV  2021-08-12  17029328  2021-08-11      17027428           1900
35083  UTICA AV  2021-08-13  17031197  2021-08-12      17029328          1869
35084  UTICA AV  2021-08-14  17032568  2021-08-13      17031197          1371
35085  UTICA AV  2021-08-15  17033846  2021-08-14      17032568          1278
35086  UTICA AV  2021-08-16  17035644  2021-08-15      17033846          1798
35087  UTICA AV  2021-08-17  17037569  2021-08-16      17035644          1925
35088  UTICA AV  2021-08-18  17039469  2021-08-17      17037569          1900
35089  UTICA AV  2021-08-19  17041353  2021-08-18      17039469          1884
35090  UTICA AV  2021-08-20  17043176  2021-08-19      17041353          1823
35091  UTICA AV  2021-08-21  17044567  2021-08-20      17043176          1391
35092  UTICA AV  2021-08-22  17045401  2021-08-21      17044567           834
35093  UTICA AV  2021-08-23  17046994  2021-08-22      17045401          1593
35094  UTICA AV  2021-08-24  17048945  2021-08-23      17046994          1951
35095  UTICA AV  2021-08-25  17050832  2021-08-24      17048945          1887
35096  UTICA AV  2021-08-26  17052835  2021-08-25      17050832          2003
35097  UTICA AV  2021-08-27  17054769  2021-08-26      17052835          1934
35098  UTICA AV  2021-08-28  17056223  2021-08-27      17054769          1454
35099  UTICA AV  2021-08-29  17057564  2021-08-28      17056223          1341
35100  UTICA AV  2021-08-30  17059443  2021-08-29      17057564          1879
35101  UTICA AV  2021-08-31  17061365  2021-08-30      17059443          1922

```

```
[70]: #ok lets check the maximum daily entries in the dataframe then assign a
      ↪threshold accordingly to reset
      #the counter to zero if we encounter a value thats even bigger than out
      ↪threshold

mask = station_df.groupby(['STATION', 'DATE'])['DAILY_ENTRIES'].max().
      ↪reset_index(). \
sort_values('DAILY_ENTRIES')

mask
```

```
[70]:
```

	STATION	DATE	DAILY_ENTRIES
812	121 ST	2021-07-24	-1942719354
816	121 ST	2021-07-28	-1942718733
17094	CANARSIE-ROCKAW	2021-08-27	-1526315613
17123	CANARSIE-ROCKAW	2021-09-25	-1526304935
21091	FLATBUSH AV-B.C	2021-08-10	-1374037426
...	...	...	...
13700	BAYCHESTER AV	2021-06-26	2064066499
18893	COURT SQ	2021-06-26	2066596158
17619	CHAMBERS ST	2021-06-26	2116123427
1274	14 ST	2021-06-26	2128673609
23494	HEWES ST	2021-06-26	2147407029

[37092 rows x 3 columns]

```
[72]: def get_daily_counts(row, max_counter):
      counter = row["ENTRIES"] - row["PREV_ENTRIES"]
      if counter < 0:
          counter = -counter
      if counter > max_counter:
          print(row["ENTRIES"], row["PREV_ENTRIES"])
          counter = min(row["ENTRIES"], row["PREV_ENTRIES"])
      if counter > max_counter:
          return 0
      return counter

station_df["DAILY_ENTRIES"] = station_df.apply(get_daily_counts, axis=1,
      ↪max_counter=1000000)
```

```
370891078 0
26181733 0
16478308 0
1681050263 0
14759168 0
50331896 0
185712452 0
1160300472 0
```

1946272192 0  
3588552 1946307906  
1946307906 3588742  
3589173 1946307906  
1946307918 3589617  
1862989314 0  
1828826573 0  
855499479 1828837016  
1828838087 855499479  
18110479 0  
991789060 0  
2128673609 0  
71309890 0  
135439844 0  
168165376 0  
117444360 168165376  
168165376 117444360  
117444362 168165376  
168165376 117444362  
118374824 0  
9411864 0  
4283666 0  
1711662138 0  
9408595 0  
1827866929 0  
53804509 0  
7603219 0  
934214846 0  
688206350 0  
4690122 0  
218418454 0  
8006802 0  
1946899327 0  
102509856 0  
1610418813 0  
1475181684 0  
1781739185 0  
1441542729 0  
654494788 0  
67422098 0  
8195885 0  
135364871 0  
3572194 0  
16852032 0  
4439476 0  
906519329 0  
1953908903 0  
2000315733 0

16256532 0  
87066800 0  
8975569 0  
5311973 0  
11770280 0  
1174478619 0  
8207442 0  
2046134325 0  
1309228109 0  
12706329 0  
120082116 0  
117440512 0  
10982525 117440512  
117440512 10983176  
1253846890 0  
5228851 0  
1569885666 0  
17863225 0  
101226828 0  
4549946 0  
12596977 0  
21726349 0  
1834129654 0  
2034035190 0  
5981788 0  
6205367 0  
14087658 0  
1925733891 0  
15436315 0  
223959537 0  
14590447 0  
119790018 0  
6578939 119823341  
119824435 6578939  
136917584 0  
335854320 0  
569526656 0  
9182669 0  
101334682 0  
7470563 0  
7868122 0  
885595278 0  
1040921584 0  
117520314 1040931990  
1398458792 0  
1208909703 0  
1843771080 0  
12510966 0

9889659 0  
100663299 0  
11591493 0  
17689146 0  
9917687 0  
13159155 0  
9983299 0  
689473925 0  
117442922 0  
8986389 0  
297605156 0  
4453577 0  
22569890 0  
569491707 0  
54982665 0  
69046754 0  
117440858 0  
83907836 0  
8526845 0  
8410935 0  
909043036 0  
37621254 909043641  
36570482 0  
4694987 0  
12129706 0  
23570953 0  
1189019447 23689857  
646572814 0  
23169825 646581434  
646583743 23169825  
3172708 0  
5969503 0  
19738774 0  
2442028 0  
67109632 0  
51149054 67109640  
5986209 0  
36569106 0  
11576096 0  
654313478 0  
1932400 0  
117440724 0  
4131859 0  
9535212 0  
100663296 0  
1179269015 0  
8934393 0  
1325333385 0

3222240 0  
1928603510 0  
8173196 0  
875171205 0  
1560439089 0  
154194177 0  
2064066499 0  
117440512 0  
2437507 0  
1354766 0  
1374029 0  
117440965 0  
12533549 117440985  
117440987 12533759  
5604268 0  
3376650 0  
1928243 0  
25799003 0  
167790157 25859746  
5294004 0  
87148383 0  
11691679 0  
5796298 0  
67436813 0  
13218335 0  
101218115 0  
5470076 0  
7089988 0  
1289818231 0  
12499188 0  
16379306 0  
1110507 0  
6034949 0  
17913288 0  
22037061 0  
990381610 0  
8836724 0  
389169819 0  
3277594 0  
6633028 0  
607993803 0  
4430067 0  
1722625870 0  
1560335770 0  
34020206 1560335819  
1560335842 34029604  
34030907 1560335842  
2043327134 0

9719309 0  
14314945 0  
3030185 0  
52178133 0  
2116123427 0  
19880920 0  
856392879 0  
118726667 0  
14008883 0  
36784604 0  
5700270 0  
151870279 0  
605276324 0  
6856435 0  
1058665572 0  
2066596158 0  
2359738 0  
6437562 0  
118013964 0  
184682778 0  
4967829 0  
1999537633 0  
14203498 0  
4259110 0  
13563384 0  
2859918 0  
5786360 0  
4740203 5822112  
5822112 4740203  
13957612 0  
9793658 0  
101260262 0  
1559807093 0  
1124135763 0  
6429367 0  
1624510434 0  
8995214 0  
2786935 0  
1383097184 0  
9072927 1383110353  
17396292 0  
23396253 0  
1762319069 0  
6785569 0  
20111054 0  
306735962 0  
11535457 0  
169817662 0

14961728 0  
6398050 0  
1662537517 0  
6305654 0  
15326183 0  
11251770 0  
20791742 0  
55426280 0  
1208844885 0  
1254630543 0  
6071491 0  
1359769875 0  
74180831 0  
8648727 0  
2147407029 0  
1915867960 0  
11677312 0  
2573274 0  
7802703 0  
5948531 7822138  
7822138 5948531  
19699486 0  
589937978 0  
1191103372 0  
8937110 0  
202177627 0  
1762020917 0  
1291015312 0  
1205603810 0  
6511520 0  
170597358 0  
8149817 0  
50331648 0  
13968508 0  
7979566 14045092  
14051313 7979671  
5929078 0  
15999749 0  
12572418 16034831  
419446441 0  
705347143 0  
352637709 705357551  
14745284 0  
5642286 0  
4829046 0  
9116301 0  
1962579921 0  
1476923212 0



1757901102 0  
7387927 0  
6940353 0  
637578981 0  
9575583 0  
5320021 0  
117441150 0  
150995161 0  
823494354 0  
4982284 0  
138584132 0  
2906407 0  
536917974 0  
6492294 0  
16810602 0  
5885626 0  
16622705 0  
5315206 0  
1664738223 0  
8659115 0  
100663296 0  
5608386 0  
8700273 0  
4035595 0  
172379101 0  
13405112 0  
5666621 0  
17983049 0  
18257223 0  
7820161 0  
337349109 0  
7503107 0  
13074257 0  
5898605 0  
7276654 0  
6731112 0  
168542433 0  
16349442 0  
10423515 0  
150997136 0  
436905968 0  
4327382 0  
6263114 0  
14435140 0  
67268813 0  
1569344992 0  
51680645 0  
19152891 0

10341292 0  
117777645 0  
1019339 0  
3772203 0  
6460441 0  
135988116 0  
100664160 0  
7744566 0  
67620014 0  
2633107 0  
13793912 0  
8339749 0  
285343137 0  
7944205 0  
123946410 0  
15367482 0  
1006930535 0  
1129213713 0  
17094812 0  
12035961 0  
84424938 0  
71543154 0  
2591316 0  
9236439 0  
1720235844 0  
1891101745 1720311340  
369309424 0  
11882758 0  
102808272 0  
17025492 102825308  
102825371 17077888  
17089093 102825371  
621873222 0  
7452282 0  
7011137 0  
14933507 0  
22552962 0  
4411627 0  
7659338 0  
168727845 0  
555471868 0  
12890876 0  
135540928 0  
5311949 0  
8403983 0  
19095274 0  
1961869 19163917  
19167678 1962037

```

7185669 19167678
19171640 7185669
12559328 0
8962516 0
1580169729 0
730966278 0
17018286 0
1266947 0

```

```
[73]: station_df['DAILY_ENTRIES'].describe()
```

```

[73]: count      37092.000000
      mean         773.418392
      std       13451.616270
      min           0.000000
      25%        174.000000
      50%        417.000000
      75%        705.000000
      max     952501.000000
      Name: DAILY_ENTRIES, dtype: float64

```

```
[75]: station_df[station_df.DAILY_ENTRIES < 0]
```

```

[75]: Empty DataFrame
      Columns: [STATION, DATE, ENTRIES, PREV_DATE, PREV_ENTRIES, DAILY_ENTRIES]
      Index: []

```

- OK! Now that we have cleaned up the dataframe let's answer the following: • What are the busiest train stations across New York City?

```
[85]: station_df
```

```

[85]:
      STATION  DATE  ENTRIES  PREV_DATE  PREV_ENTRIES  DAILY_ENTRIES
0      1 AV 2021-06-26  370891078  1970-01-01           0           0
1      1 AV 2021-06-27  370891152  2021-06-26  370891078           74
2      1 AV 2021-06-28  370891228  2021-06-27  370891152           76
3      1 AV 2021-06-29  370891317  2021-06-28  370891228           89
4      1 AV 2021-06-30  370891419  2021-06-29  370891317          102
...
37087  ZEREGA AV 2021-09-27  1308054  2021-09-26  1307428          626
37088  ZEREGA AV 2021-09-28  1308726  2021-09-27  1308054          672
37089  ZEREGA AV 2021-09-29  1309387  2021-09-28  1308726          661
37090  ZEREGA AV 2021-09-30  1310068  2021-09-29  1309387          681
37091  ZEREGA AV 2021-10-01  1310763  2021-09-30  1310068          695

```

```
[37092 rows x 6 columns]
```

```
[113]: station_df['DATE'] = pd.to_datetime(station_df['DATE'], errors='coerce')
station_df.dtypes
```

```
[113]: STATION          object
DATE              datetime64[ns]
ENTRIES           int64
PREV_DATE         datetime64[ns]
PREV_ENTRIES       int64
DAILY_ENTRIES      int64
dtype: object
```

```
[114]: station_df['weekNumber'] = station_df['DATE'].dt.week
```

```
/var/folders/4y/plz6nn617g3gccj1l9_90bhc0000gn/T/ipykernel_3542/3323928275.py:1:
FutureWarning: Series.dt.weekofyear and Series.dt.week have been deprecated.
Please use Series.dt.isocalendar().week instead.
station_df['weekNumber'] = station_df['DATE'].dt.week
```

```
[115]: station_df
```

```
[115]:
```

	STATION	DATE	ENTRIES	PREV_DATE	PREV_ENTRIES	\
0	1 AV	2021-06-26	370891078	1970-01-01	0	
1	1 AV	2021-06-27	370891152	2021-06-26	370891078	
2	1 AV	2021-06-28	370891228	2021-06-27	370891152	
3	1 AV	2021-06-29	370891317	2021-06-28	370891228	
4	1 AV	2021-06-30	370891419	2021-06-29	370891317	
...	...	...	...	...	...	
37087	ZEREGA AV	2021-09-27	1308054	2021-09-26	1307428	
37088	ZEREGA AV	2021-09-28	1308726	2021-09-27	1308054	
37089	ZEREGA AV	2021-09-29	1309387	2021-09-28	1308726	
37090	ZEREGA AV	2021-09-30	1310068	2021-09-29	1309387	
37091	ZEREGA AV	2021-10-01	1310763	2021-09-30	1310068	
	DAILY_ENTRIES	weekNumber				
0	0	25				
1	74	25				
2	76	26				
3	89	26				
4	102	26				
...	...	...				
37087	626	39				
37088	672	39				
37089	661	39				
37090	681	39				
37091	695	39				

```
[37092 rows x 7 columns]
```

```
[130]: busiest_station = station_df.groupby(['weekNumber', 'STATION'])['DAILY_ENTRIES'].
      ↪sum().reset_index().\
      sort_values('DAILY_ENTRIES', ascending = False)
      #df.groupby(month('date')).agg({'Revenue': 'sum'})
      #b.groupby(by=[b.index.month, b.index.year])
```

```
[131]: busiest_station
```

```
[131]:
```

	weekNumber	STATION	DAILY_ENTRIES
258	25	JOURNAL SQUARE	953564
181	25	CHAUNCEY ST	932938
235	25	GROVE STREET	822180
256	25	JFK JAMAICA CT1	751924
184	25	CITY / BUS	738037
...	...	...	...
1470	28	ROCKAWAY PARK B	0
5446	39	BEACH 105 ST	0
4952	38	170 ST	0
405	26	170 ST	0
3293	33	KINGS HWY	0

[5684 rows x 3 columns]

```
[138]: test1 = busiest_station.groupby(['STATION'])['DAILY_ENTRIES'].mean().
      ↪reset_index().\
      sort_values('DAILY_ENTRIES', ascending = False).head(10)
```

```
[139]: test1
```

```
[139]:
```

	STATION	DAILY_ENTRIES
256	JFK JAMAICA CT1	74714.466667
258	JOURNAL SQUARE	72213.733333
181	CHAUNCEY ST	65311.933333
235	GROVE STREET	62278.066667
184	CITY / BUS	54328.333333
351	THIRTY ST	53497.200000
298	NEWARK BM BW	46346.066667
352	THIRTY THIRD ST	46237.466667
354	TOMPKINSVILLE	45850.533333
315	PATH NEW WTC	40581.733333

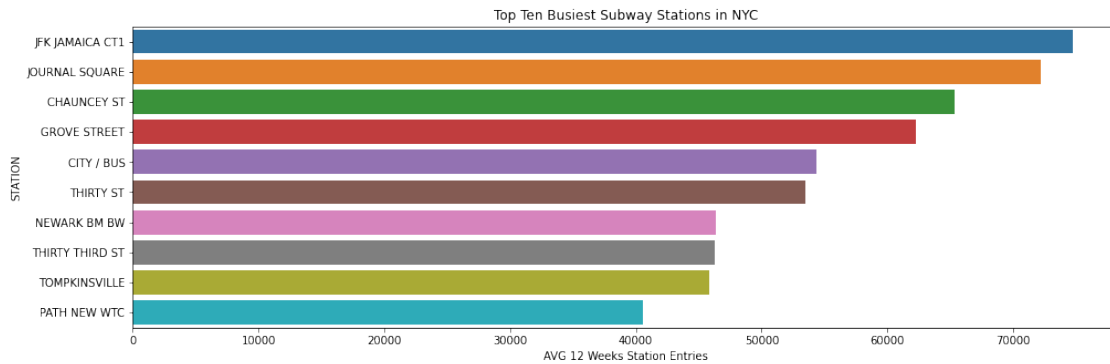
```
[157]: %matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns

fig_dims = (16, 5)
```

```
fig, ax = plt.subplots(figsize=fig_dims)

my_plot = sns.barplot(x='DAILY_ENTRIES', y='STATION', ax=ax, data=test1)
my_plot.set(xlabel="AVG 12 Weeks Station Entries", title='Top Ten Busiest Subway Stations in NYC')
```

```
[157]: [Text(0.5, 0, 'AVG 12 Weeks Station Entries'),
Text(0.5, 1.0, 'Top Ten Busiest Subway Stations in NYC')]
```



```
[ ]:
```