

# Data-Driven Decision Making - Assignment Project

201306-citibike-tripdata file contains a descriptive analysis of the CitiBike company in New York in the period of June 2013. The file contains columns:

- Trip duration in seconds (how long the bike trip lasted from start to finish)
- Start time (the date and time of day where the trip started)
- Stop time (the date and time of day where the trip lasted)
- Start station ID (the ID number of the station which the trip started)
- Start station name (the name of the station which the trip started)
- Start station latitude (value of the vertical (north/south) axis in the coordinate plane, used in maps to locate)
- Start station longitude (value of horizontal (east/west) axis in the coordinate plane, used in maps to locate)
- End station ID (the ID number of the station which the trip ended)
- End station name (the name of the station which the trip ended)
- End station latitude (value of the vertical (north/south) axis in the coordinate plane, used in maps to locate)
- End station longitude (value of horizontal (east/west) axis in the coordinate plane, used in maps to locate)
- Bike ID (the ID of the bike that was used)
- User Type (if the person using the bike is a customer or a subscriber/annual member)
- Birth Year (the birthdate of the person using the bike)
- Gender (the gender of the person using the bike)

Each row in the data represents a single use of the bike from start to finish and the description of the user (the person that rents the bikes).

## Missing values (NULL):

There were some missing values in the data (in the thousands!) that had the value “NULL”. I changed the value “NULL” and replaced it with “Unknown”, using ctrl+H:

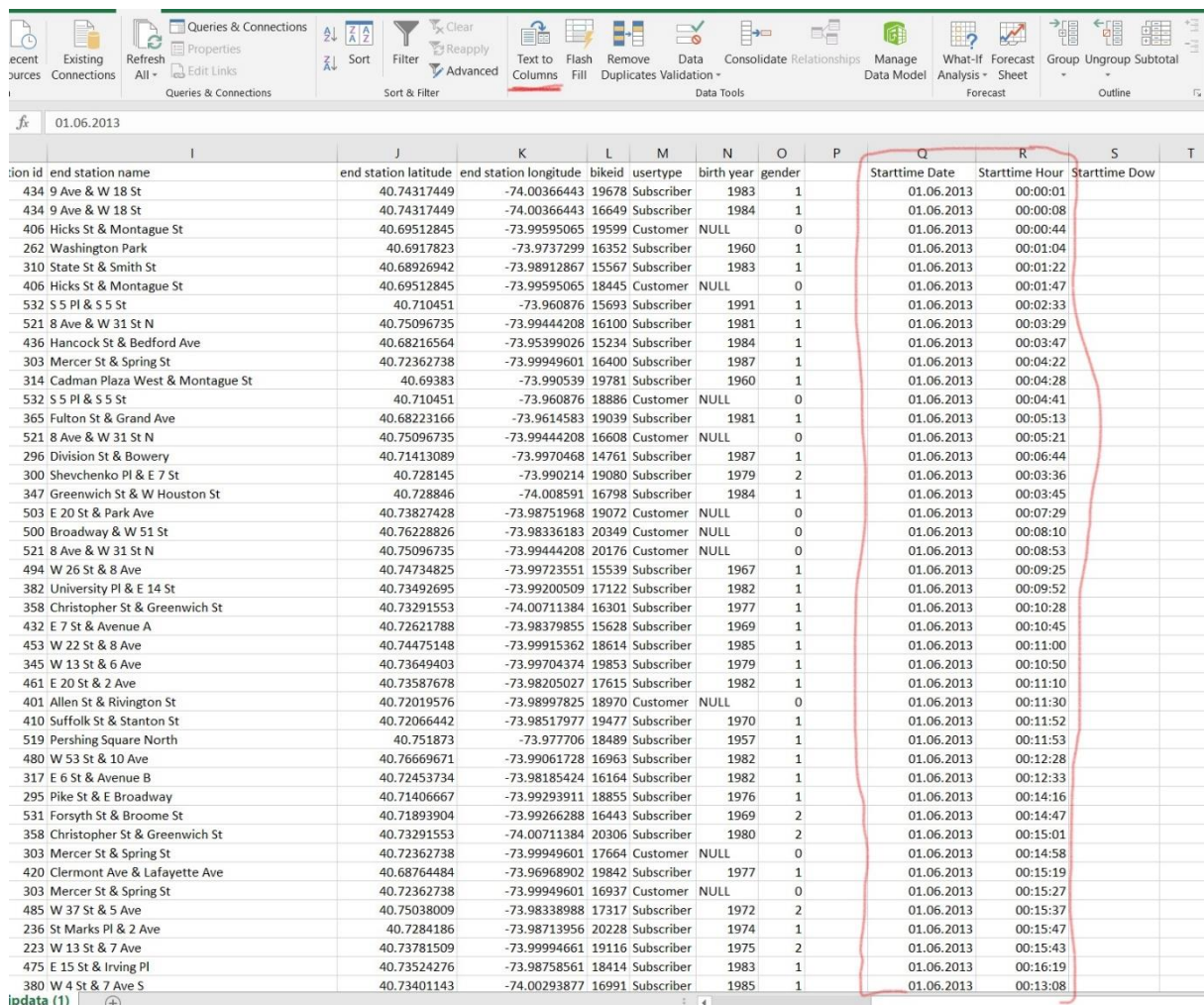
40.68981035	-73.97493121	365	Fulton St & Grand Ave	40.6822316
40.7462009	-73.98855723	521	8 Ave & W 31 St N	40.7509673
40.744023	-73.976056	296	Division St & Bowery	40.7141308
40.7149787	-74.013012	300	Shevchenko Pl & E 7 St	40.72814
40.71497				40.72884
40.738274				40.7382742
40.719392				40.7622882
40.74620				40.7509673
40.747348				40.7473482
40.734011				40.7349269
40.760957				40.7329155
40.766696				40.7262178
40.744751				40.7447514
40.737815				40.7364940
40.743943				40.7358767
40.722103				40.7201957
40.7220				40.7206644
40.74734825	-73.99723551	519	Pershing Square North	40.75187
40.73935542	-73.99931783	480	W 53 St & 10 Ave	40.7666967
40.7262807	-73.98978041	317	E 6 St & Avenue B	40.7245373
40.71893904	-73.99266288	295	Pike St & E Broadway	40.7140666
40.71893904	-73.99266288	531	Forsyth St & Broome St	40.7189390
40.73291553	-74.00711384	358	Christopher St & Greenwich St	40.7329155
40.71603118	-73.99743371	303	Market St & Spring St	40.7160312

Figure 1: Replacing “NULL” with “Unknown”.

Why I choose to have the missing values is because I believe the missing values is still a value that can be analyzed. What I mean by that is that I found no missing values or “NULL” in the “start station name” column, but a lot in the “end station name” column (also in other columns like end station latitude/longitude). It could mean that rented bikes from start stations did not end in an end station, but rather in unknown places in New York city, where people just ended their biketrip in unknown locations. Or people just rented a bike and never actually go nowhere. It could be a lot of reasons. Maybe keeping the missing values will make the analysis difficult. Maybe I have to delete the missing values. I will try to figure this out in this assignment. So long KISS (Keep it simple stupid) ☺

### Assignment 1:

To create three new variables (columns), the first thing that I did was to use “text to columns” inside the “data” tab to split column B, since column B has both the date and time in the same column. (Note that before I used “text to columns” I copied column B and pasted it in column Q). Figure 1 below shows the result.



ion id	end station name	end station latitude	end station longitude	bikeid	usertype	birth year	gender	Starttime Date	Starttime Hour	Starttime Dow
434	9 Ave & W 18 St	40.74317449	-74.00366443	19678	Subscriber	1983	1	01.06.2013	00:00:01	
434	9 Ave & W 18 St	40.74317449	-74.00366443	16649	Subscriber	1984	1	01.06.2013	00:00:08	
406	Hicks St & Montague St	40.69512845	-73.99595065	19599	Customer	NULL	0	01.06.2013	00:00:44	
262	Washington Park	40.6917823	-73.9737299	16352	Subscriber	1960	1	01.06.2013	00:01:04	
310	State St & Smith St	40.68926942	-73.98912867	15567	Subscriber	1983	1	01.06.2013	00:01:22	
406	Hicks St & Montague St	40.69512845	-73.99595065	18445	Customer	NULL	0	01.06.2013	00:01:47	
532	S 5 Pl & S 5 St	40.710451	-73.960876	15693	Subscriber	1991	1	01.06.2013	00:02:33	
521	8 Ave & W 31 St N	40.75096735	-73.99444208	16100	Subscriber	1981	1	01.06.2013	00:03:29	
436	Hancock St & Bedford Ave	40.68216564	-73.95399026	15234	Subscriber	1984	1	01.06.2013	00:03:47	
303	Mercer St & Spring St	40.72362738	-73.99949601	16400	Subscriber	1987	1	01.06.2013	00:04:22	
314	Cadman Plaza West & Montague St	40.69383	-73.990539	19781	Subscriber	1960	1	01.06.2013	00:04:28	
532	S 5 Pl & S 5 St	40.710451	-73.960876	18886	Customer	NULL	0	01.06.2013	00:04:41	
365	Fulton St & Grand Ave	40.68223166	-73.9614583	19039	Subscriber	1981	1	01.06.2013	00:05:13	
521	8 Ave & W 31 St N	40.75096735	-73.99444208	16608	Customer	NULL	0	01.06.2013	00:05:21	
296	Division St & Bowery	40.71413089	-73.9970468	14761	Subscriber	1987	1	01.06.2013	00:06:44	
300	Shevchenko Pl & E 7 St	40.728145	-73.990214	19080	Subscriber	1979	2	01.06.2013	00:03:36	
347	Greenwich St & W Houston St	40.728846	-74.008591	16798	Subscriber	1984	1	01.06.2013	00:03:45	
503	E 20 St & Park Ave	40.73827428	-73.98751968	19072	Customer	NULL	0	01.06.2013	00:07:29	
500	Broadway & W 51 St	40.76228826	-73.98336183	20349	Customer	NULL	0	01.06.2013	00:08:10	
521	8 Ave & W 31 St N	40.75096735	-73.99444208	20176	Customer	NULL	0	01.06.2013	00:08:53	
494	W 26 St & 8 Ave	40.74734825	-73.99723551	15539	Subscriber	1967	1	01.06.2013	00:09:25	
382	University Pl & E 14 St	40.73492695	-73.99200509	17122	Subscriber	1982	1	01.06.2013	00:09:52	
358	Christopher St & Greenwich St	40.73291553	-74.00711384	16301	Subscriber	1977	1	01.06.2013	00:10:28	
432	E 7 St & Avenue A	40.72621788	-73.98379855	15628	Subscriber	1969	1	01.06.2013	00:10:45	
453	W 22 St & 8 Ave	40.74475148	-73.99915362	18614	Subscriber	1985	1	01.06.2013	00:11:00	
345	W 13 St & 6 Ave	40.73649403	-73.99704374	19853	Subscriber	1979	1	01.06.2013	00:10:50	
461	E 20 St & 2 Ave	40.73587678	-73.98205027	17615	Subscriber	1982	1	01.06.2013	00:11:10	
401	Allen St & Rivington St	40.72019576	-73.98997825	18970	Customer	NULL	0	01.06.2013	00:11:30	
410	Suffolk St & Stanton St	40.72066442	-73.98517977	19477	Subscriber	1970	1	01.06.2013	00:11:52	
519	Pershing Square North	40.751873	-73.977706	18489	Subscriber	1957	1	01.06.2013	00:11:53	
480	W 53 St & 10 Ave	40.76669671	-73.99061728	16963	Subscriber	1982	1	01.06.2013	00:12:28	
317	E 6 St & Avenue B	40.72453734	-73.98185424	16164	Subscriber	1982	1	01.06.2013	00:12:33	
295	Pike St & E Broadway	40.71406667	-73.99293911	18855	Subscriber	1976	1	01.06.2013	00:14:16	
531	Forsyth St & Broome St	40.71893904	-73.99266288	16443	Subscriber	1969	2	01.06.2013	00:14:47	
358	Christopher St & Greenwich St	40.73291553	-74.00711384	20306	Subscriber	1980	2	01.06.2013	00:15:01	
303	Mercer St & Spring St	40.72362738	-73.99949601	17664	Customer	NULL	0	01.06.2013	00:14:58	
420	Clermont Ave & Lafayette Ave	40.68764484	-73.96968902	19842	Subscriber	1977	1	01.06.2013	00:15:19	
303	Mercer St & Spring St	40.72362738	-73.99949601	16937	Customer	NULL	0	01.06.2013	00:15:27	
485	W 37 St & 5 Ave	40.75038009	-73.98338988	17317	Subscriber	1972	2	01.06.2013	00:15:37	
236	St Marks Pl & 2 Ave	40.7284186	-73.98713956	20228	Subscriber	1974	1	01.06.2013	00:15:47	
223	W 13 St & 7 Ave	40.73781509	-73.99994661	19116	Subscriber	1975	2	01.06.2013	00:15:43	
475	E 15 St & Irving Pl	40.73524276	-73.98758561	18414	Subscriber	1983	1	01.06.2013	00:16:19	
380	W 4 St & 7 Ave S	40.73401143	-74.00293877	16991	Subscriber	1985	1	01.06.2013	00:13:08	

Figure 2: Making columns “starttime date” and “starttime hour”.

To make the last column, column S “starttime dow” I used formula =weekday and added column Q in the formula. I got number 7, which means Saturday (in excel 1 is from Sunday to 7 that is Saturday, as

you probably know). I would like to have the text of a day instead of numbers, so I used “format cells”, then custom inside category and typed dddd, as shown in the next page.

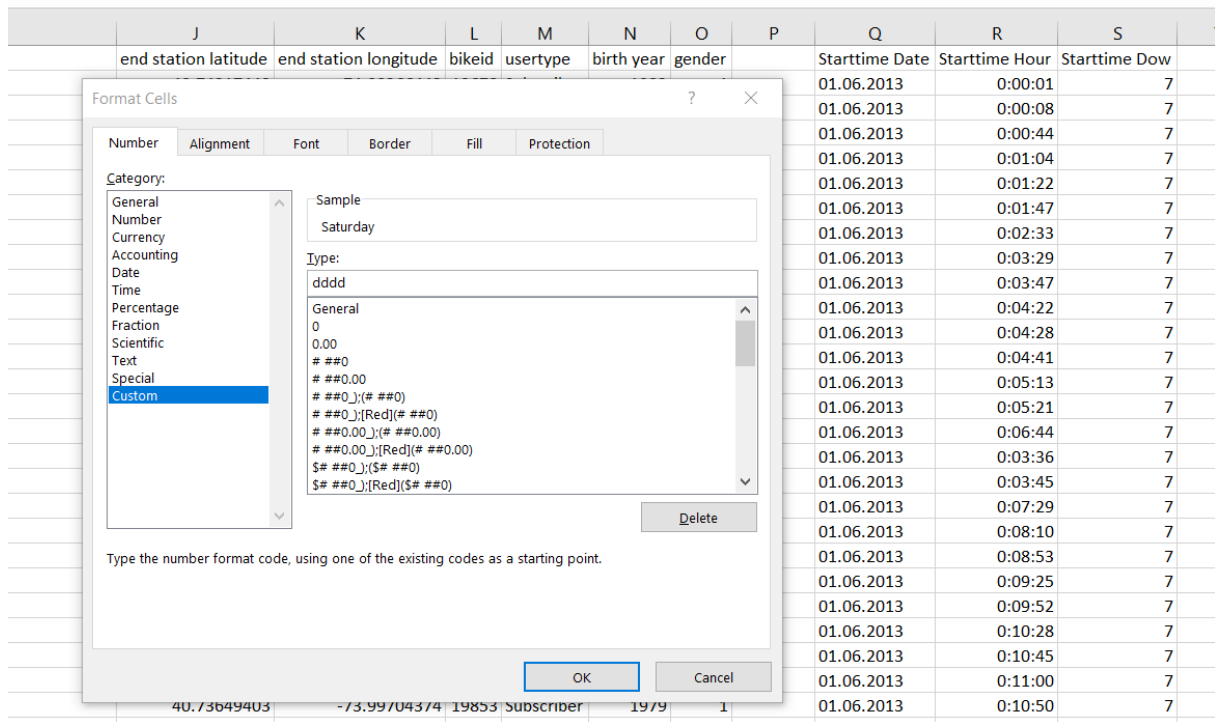


Figure 3: Using “format cells” to replace 7 to Saturday.

	M	N	O	P	Q	R	S	T
id	usertype	birth year	gender		Starttime Date	Starttime Hour	Starttime Dow	
78	Subscriber	1983	1		01.06.2013	0:00:01	Saturday	
49	Subscriber	1984	1		01.06.2013	0:00:08	Saturday	
99	Customer	NULL	0		01.06.2013	0:00:44	Saturday	
52	Subscriber	1960	1		01.06.2013	0:01:04	Saturday	
67	Subscriber	1983	1		01.06.2013	0:01:22	Saturday	
45	Customer	NULL	0		01.06.2013	0:01:47	Saturday	
93	Subscriber	1991	1		01.06.2013	0:02:33	Saturday	
30	Subscriber	1981	1		01.06.2013	0:03:29	Saturday	
34	Subscriber	1984	1		01.06.2013	0:03:47	Saturday	
00	Subscriber	1987	1		01.06.2013	0:04:22	Saturday	
81	Subscriber	1960	1		01.06.2013	0:04:28	Saturday	
86	Customer	NULL	0		01.06.2013	0:04:41	Saturday	
39	Subscriber	1981	1		01.06.2013	0:05:13	Saturday	
08	Customer	NULL	0		01.06.2013	0:05:21	Saturday	
61	Subscriber	1987	1		01.06.2013	0:06:44	Saturday	
80	Subscriber	1979	2		01.06.2013	0:03:36	Saturday	
98	Subscriber	1984	1		01.06.2013	0:03:45	Saturday	
72	Customer	NULL	0		01.06.2013	0:07:29	Saturday	
49	Customer	NULL	0		01.06.2013	0:08:10	Saturday	
76	Customer	NULL	0		01.06.2013	0:08:53	Saturday	
39	Subscriber	1967	1		01.06.2013	0:09:25	Saturday	
22	Subscriber	1982	1		01.06.2013	0:09:52	Saturday	
01	Subscriber	1977	1		01.06.2013	0:10:28	Saturday	
28	Subscriber	1969	1		01.06.2013	0:10:45	Saturday	
14	Subscriber	1985	1		01.06.2013	0:11:00	Saturday	
53	Subscriber	1979	1		01.06.2013	0:10:50	Saturday	
15	Subscriber	1982	1		01.06.2013	0:11:10	Saturday	
70	Customer	NULL	0		01.06.2013	0:11:30	Saturday	
77	Subscriber	1970	1		01.06.2013	0:11:52	Saturday	
89	Subscriber	1957	1		01.06.2013	0:11:53	Saturday	
63	Subscriber	1982	1		01.06.2013	0:12:28	Saturday	
64	Subscriber	1982	1		01.06.2013	0:12:33	Saturday	
55	Subscriber	1976	1		01.06.2013	0:14:16	Saturday	
43	Subscriber	1969	2		01.06.2013	0:14:47	Saturday	
06	Subscriber	1980	2		01.06.2013	0:15:01	Saturday	
64	Customer	NULL	0		01.06.2013	0:14:58	Saturday	

Figure 4: Saturday instead of 7.

### 24hour clock:

The hour on both column “starttime hour” and “stoptime hour” I chose to use a 24hour clock instead of using a.m. and p.m. I tried to use a.m. and p.m., but my excel for some reason got stuck every time I tried to use a.m./p.m., and my whole sheet got weird and just kept loading. But from the information on Kaggle, it seems that they used a 24hour clock on their data.

### Assignment 2:

To look at which day of the week rentals are most popular and least popular, I used a chart to visualize a total of bike rentals from 01.06.2013 to 01.07.2013. To find the values, I used =COUNTIF.

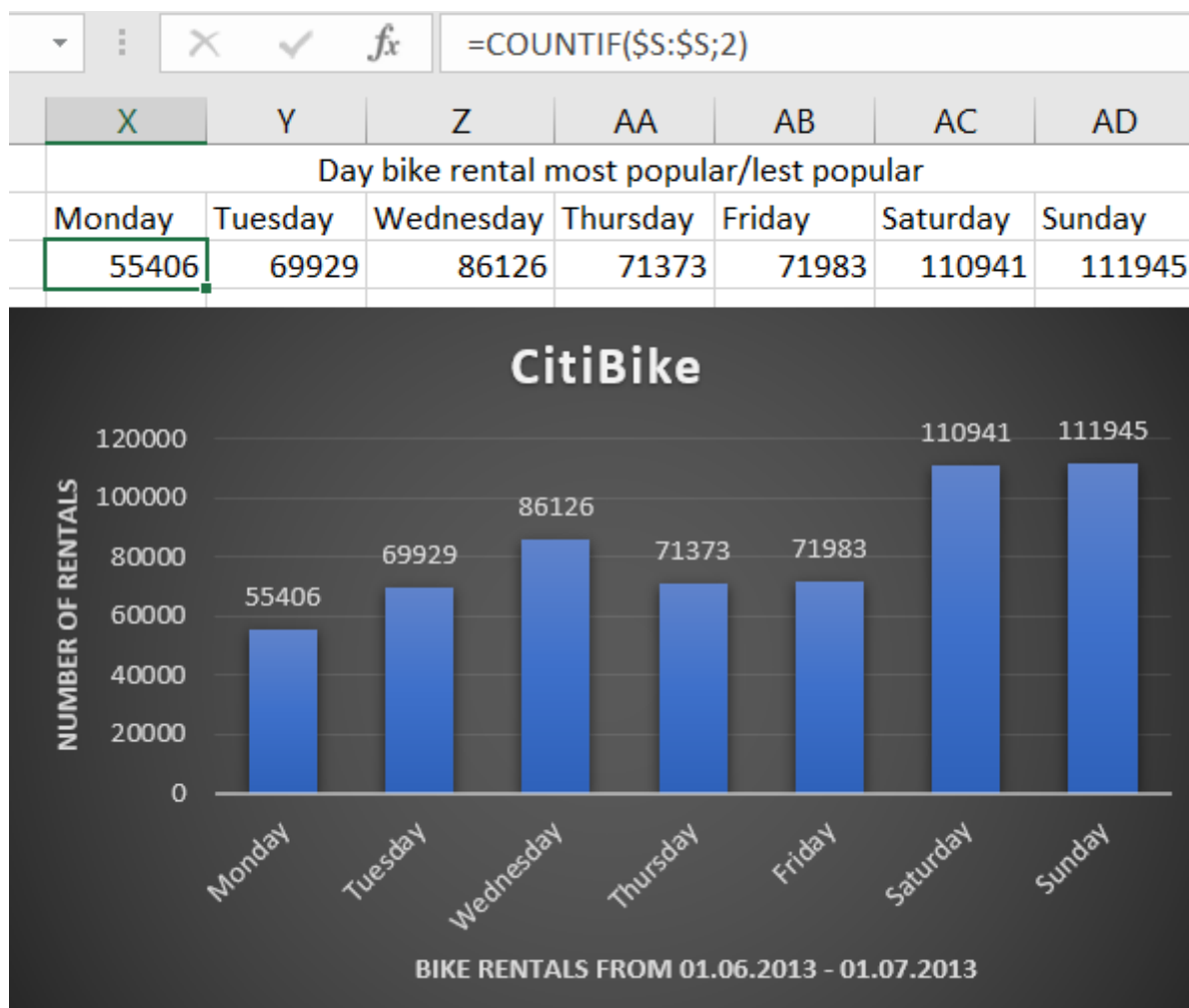


Figure 5: Rentals most popular/ least popular from 01.06.13 to 01.07.13.

As the chart shows:

- Most popular days for renting a bike is in the weekends, Saturdays and Sundays. It could be because people regularly have the day off from work/school, and that people has more time to do recreational purposes or visiting and meeting friends and family.
- Least popular days for renting a bike is in the start of a week, Mondays and Tuesdays. It could be because people are at work or school. It could also mean that people has done

recreational activities in the weekend and is likely not that active on Mondays and Tuesdays. They rather want to stay at home.

The average number of rented bikes on weekends compare to the average number of rented bikes during working days is shown below by using formula =average. The average number from Monday to Friday is 70963 rented bikes. On weekends from Saturday to Sunday the average rented bikes are 111443.

- Though there are only two days in a weekend, the average amount of people renting bikes is still higher than in working days which has five days total.

Average number of rented bikes on working days
70963.4
Average number of rented bikes on weekends
111443

Figure 6: Compared average rented bikes on weekends (Saturdays to Sundays) vs. working days (from Mondays to Fridays)

Now, which hour in a day is the most popular and the least popular time to rent a bike in New York City from 01.06.13 to 01.07.13? To find that out, I created a chart using bins with a duration of 2 hours between each graph (except the first and last hour in a day). I used formula =FREQUENCY(\$R\$2:\$R\$577704;AB22:AB34).

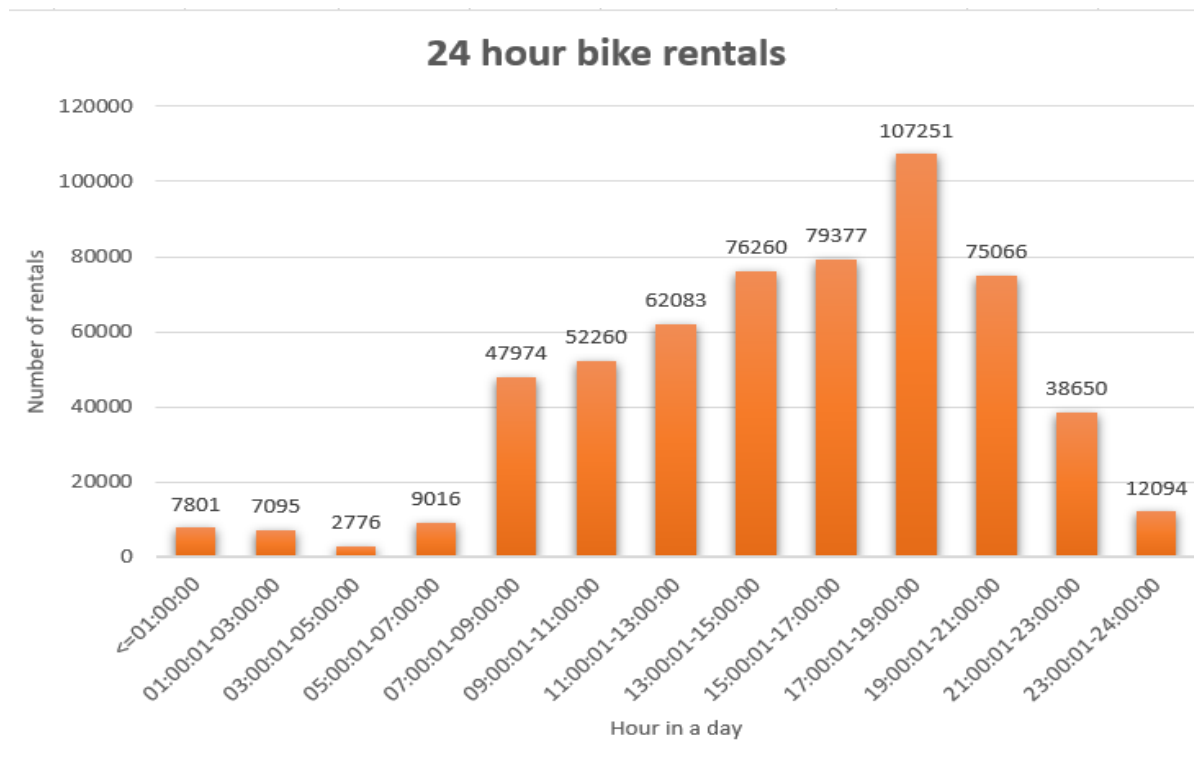


Figure 7: Showing time of day rental bikes are used.

The chart shows:

- The most popular time people renting bikes was between 17:00:01 to 19:00:00 in the evening with a total of 107251 rental bikes. This could be the time when people are finish at work to either travel home or meet family and friend etc.
- The least popular time was between 03:00:01 to 05:00:00 in the morning with a total of 2776 rental bikes. This explanation could be that people normally sleeps at that time of day.

The most popular start stations for bike rentals is shown below. I used a PivotTable to find the top 10 most popular start stations.

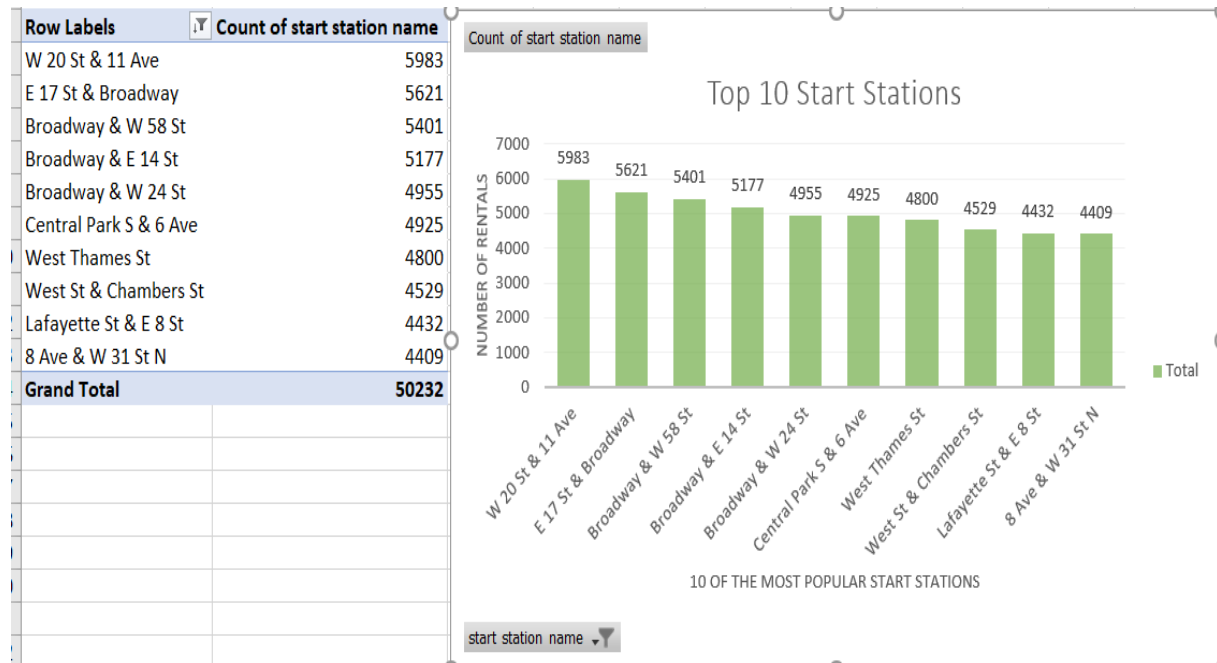


Figure 8: Top 10 start stations



The least popular start stations for bike rentals is also shown below.

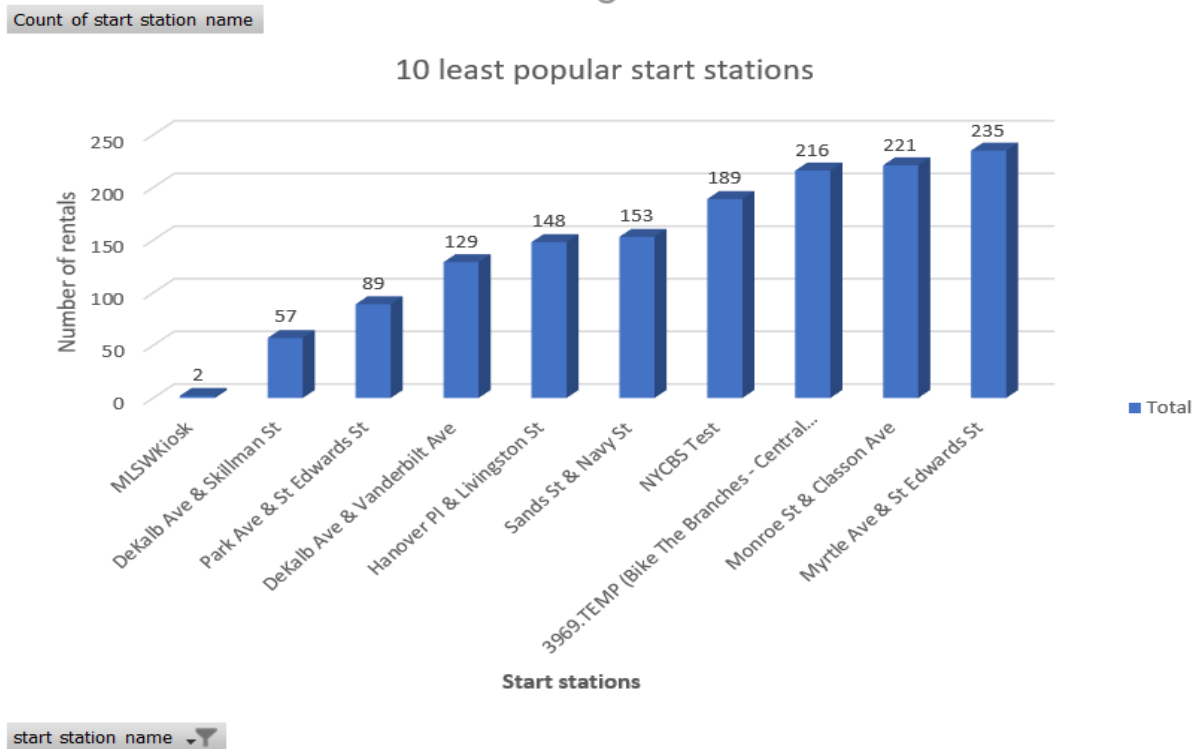


Figure 9: 10 least popular start stations for bike rentals

Figure 8 and figure 9 shows:

- The most popular start station to rent a bike was at West 20 Street & 11 Avenue with 5983 between 01.06.13 to 01.07.13.
- The least popular start station to rent a bike was at MLSWKiosk with only 2 bikes between 01.06.13 to 01.07.13.

It's quite the difference between the most and the least. Maybe the area in West 20 Street & 11 Avenue is a well-known station to rent a bike. Maybe it's located in an area where other transportations meet, like a bus station or a boat pier. It could be a lot of reasons. Why the MLSWKiosk has a low rental result in that same period could be that the location is at an area where most people either rides the bus, subway, car etc. Maybe the kiosk is located in a far-off area.

The most popular destinations for people how rented bikes are shown below:

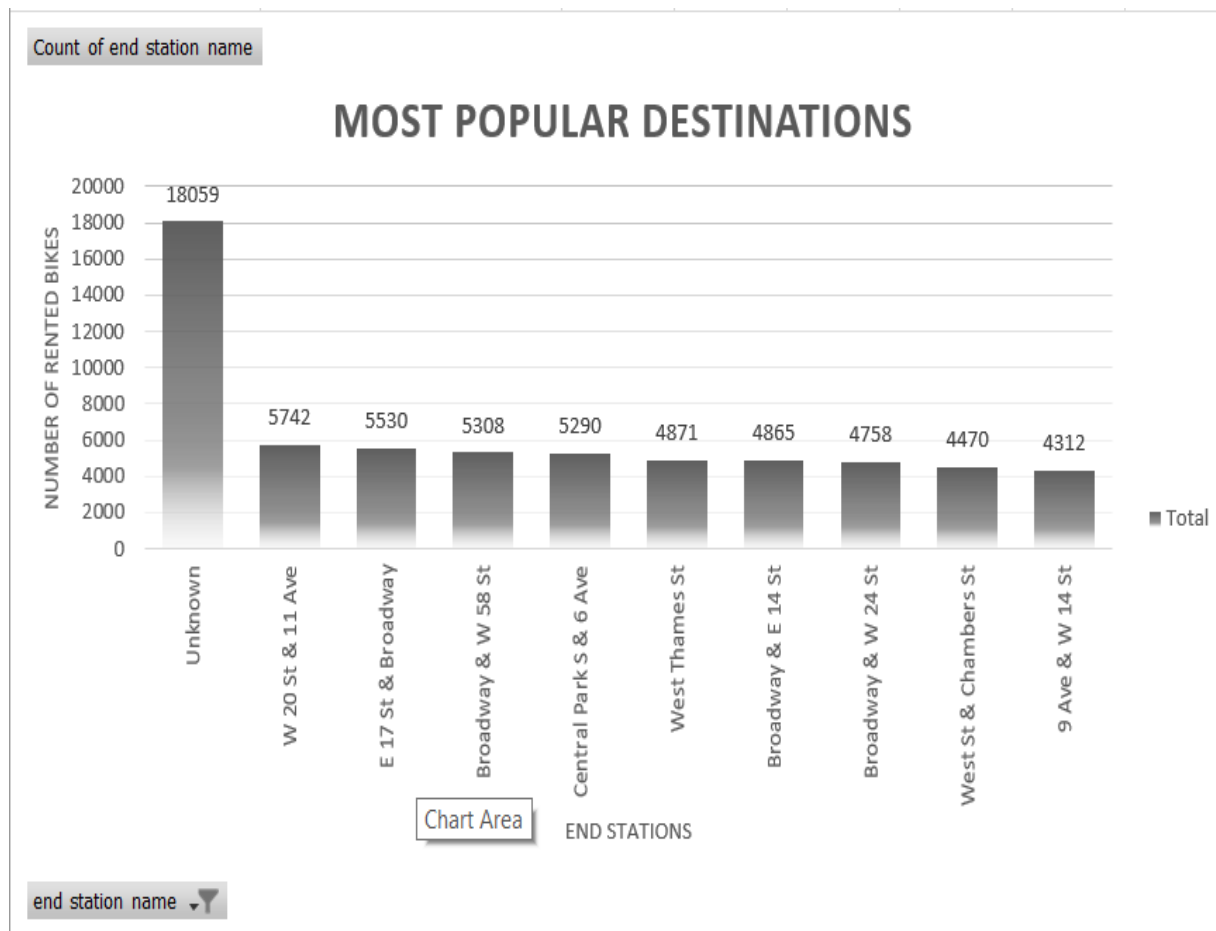


Figure 10: Most popular destination for rental bikes

What the chart shows:

- Destination “Unknown” shows, by far, the most frequent value in the chart. It can be that most of the time people how rents bikes don’t travel to an end station, but rather leave the bikes in unknown places. I hope CitiBike in New York had some kind of tracking devices on their bikes in 2013. If not, this could be costly business to run if 18059 bikes went missing!
- The most popular destination, or end station was West 20 street & 11 Avenue with 5742 rented bikes. It’s the most popular destination and the most popular start station (as shown in figure 10).
- The second most popular destination was East 17 & Broadway with 5530 rented bikes. The same address was also the second most popular start station (as shown in figure 10).
- As I said earlier, the reason why I chose to have the missing values in the data “unknown” is because it maybe has a logical meaning.



The least popular destination is shown below:

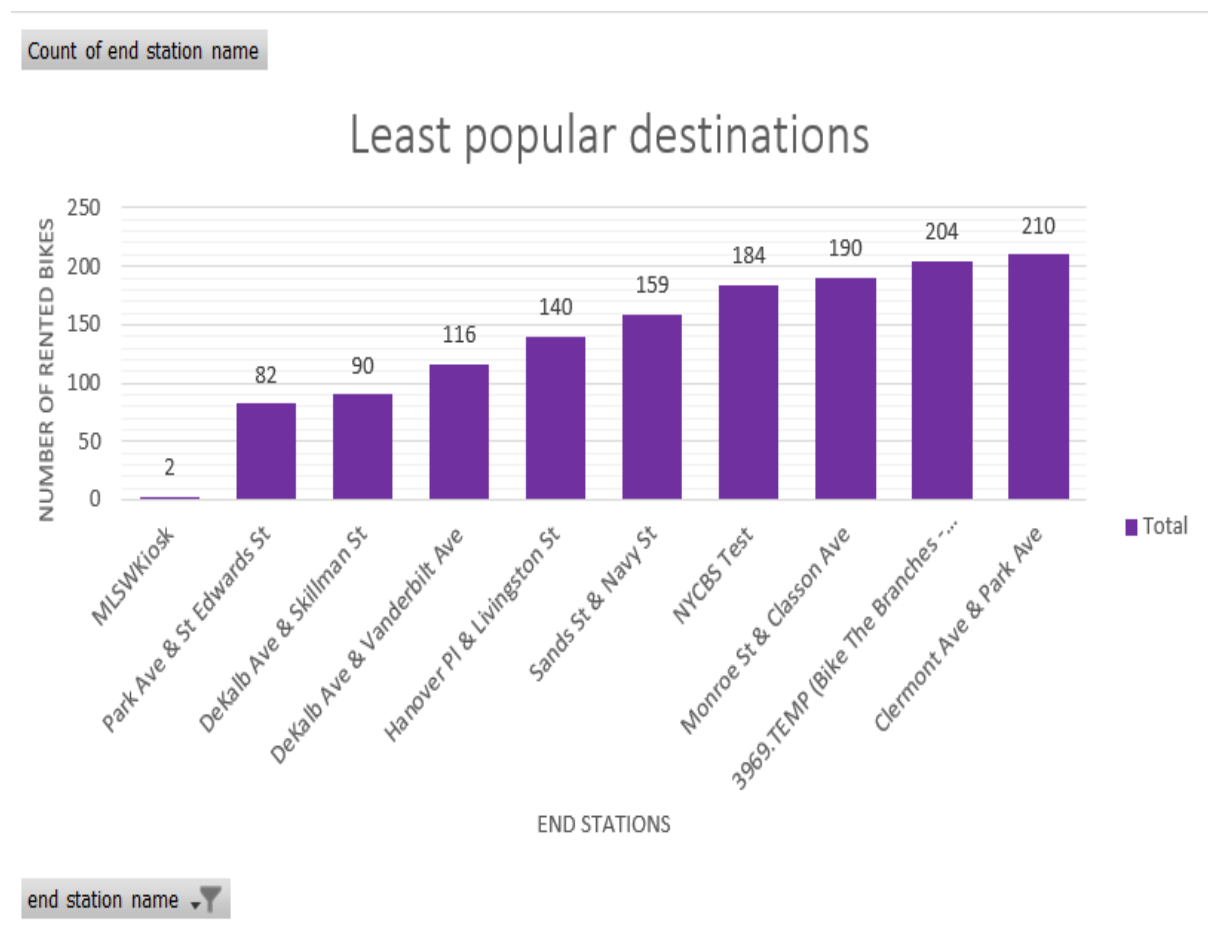


Figure 11: Top 10 least popular destinations between 01.06.13 to 01.07.13.

What the chart shows:

- MLSWKiosk was the least popular destination for people renting bikes with a value of 2. Like in figure 9, MLSWKiosk was the least popular start station with also 2 bikes from 01.06.13 to 01.07.13.
- Park Ave & St. Edwards Street was the second least popular destination, followed by DeKalb Avenue & Skillman Street. The same address was the second least popular start station to rent a bike, as shown in figure 10.

Are start stations and end stations usually the same or different?

That question was hard to figure out. After getting a tip from a fellow student in Discord, I used the station IDs to see if the ID numbers starts and ends matches between the same rows on the sheet. I'm not quite sure if I understood the question completely, but I figured out a solution:

- To find both matching values and not matching values in the same rows I used the =IF function as shown below.



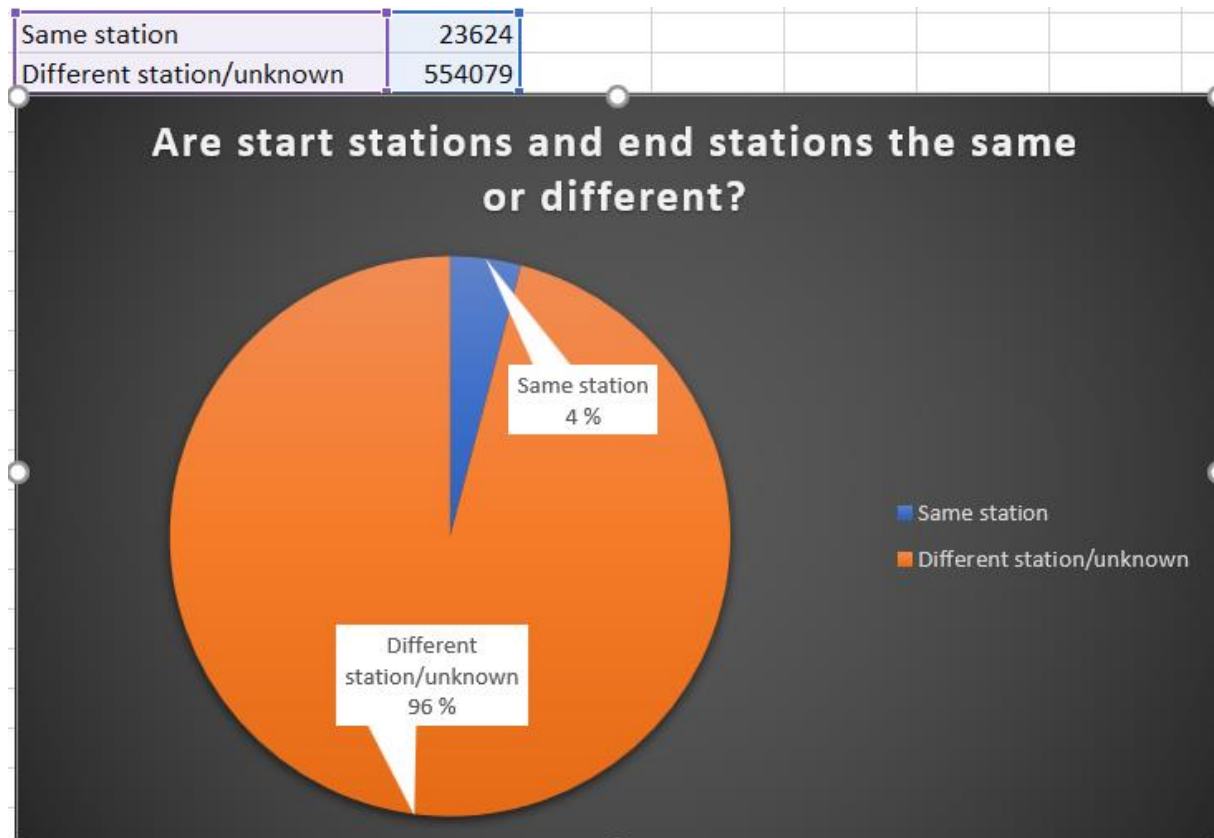
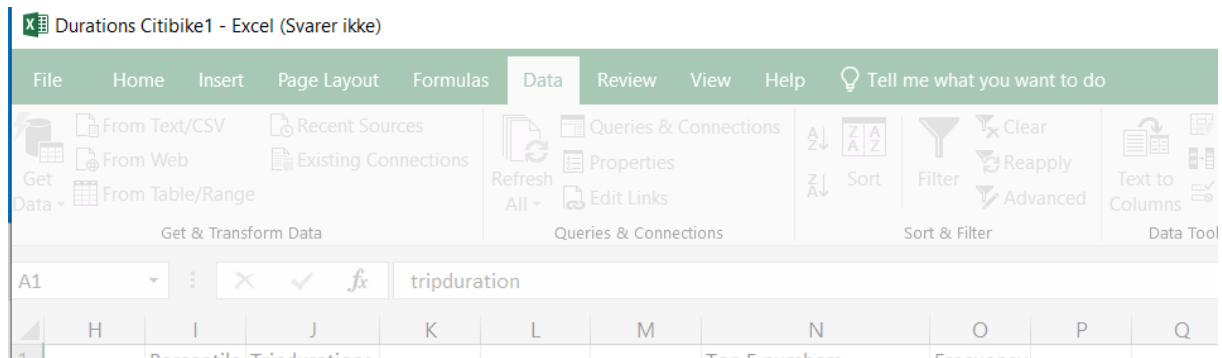


Figure 14: Showing if the destination for people renting bikes ended at the same place as the starting station, or if the destination ended in a different station/unknown location in %.

What the pie chart is showing:

- 96 % of the people who rented bikes usually did not end their destination in the same station where they started from. The bikes ended in other stations or in an unknown area.
- Only 4 % of trips with rented bikes ended at the same station where the bikes started from.

**Note:** "In this point of assignment my excel started to load a lot and sometimes crashed. I tried to delete all rows and columns that where empty, even changed to a new workbook and copy/paste single columns for analysis, but it kept crashing. Instead of copying charts directly from excel to word, I took a screenshot of every chart and copied it in word. I spent hours of valuable time waiting, fixing, finding ways to make excel run better. Sometimes I had to do calculations all over again. I also had a hard time using the Data Analysis Toolpak because it just kept loading and crashing when I tried to find some results of the data using the tool".



This “scene” (picture above) became more and more frequent.

### **Assignment 3:**

By the help of the Data Analysis Toolpak in excel I got (finally....) some important values to use in this assignment:

C	D
<i>tripduration</i>	
Mean	1372.570279
Standard Error	11.87797129
Median	874
Mode	547
Standard Deviation	9028.063033
Sample Variance	81505922.12
Kurtosis	78921.68157
Skewness	224.0203653
Range	3876418
Minimum	61
Maximum	3876479
Sum	792937968
Count	577703

Figure 15: Using the Data Analysis Toolpak in excel.

- The values calculated by the Toolpak already shows me that the values inside column “trip durations” has some extreme variations.
- Sample variance has a variability of 81505922. This tells me the degree of spread in the data. The variance is so different in the relation to the mean value of 1372,57.
- The standard deviation is quite high. High standard deviation means that the values in the column are spread out.
- The range between minimum and maximum has a range of 3876418.

- Skewness looks like its skewed to the right/positive.
- As I can remember, if the mean is greater than the median, the distribution of the “trip durations” column has to be skewed positively/right side.
- The most frequent value in the column is 547. This could mean that most rented bike duration has a duration of 547 seconds. What I mean about that is that people who rent bikes has an average use of 547 seconds. Said differently, the average trip lasts about 9 minutes.

It looks like the data shows some extreme outliers. I think it would be difficult to make a histogram based on the extreme outliers in the “trip durations” column.

### Outliers:

Investigating on the outliers, I checked the “Custom Sort” and got this information:

Q	R	S	U	V
tripduration	Starttime Date	Starttime Hour	Stoptime Date	Stoptime Hour
3876479	04.06.2013	04:51:57	19.07.2013	01:39:56
2742506	02.06.2013	13:11:18	04.07.2013	06:59:44
1860069	19.06.2013	12:32:51	11.07.2013	01:14:00
1331178	05.06.2013	11:42:36	20.06.2013	21:28:54
1239509	23.06.2013	20:36:50	08.07.2013	04:55:19
1048424	14.06.2013	15:40:53	26.06.2013	18:54:37
865531	19.06.2013	18:49:01	29.06.2013	19:14:32
764344	12.06.2013	14:47:54	21.06.2013	11:06:58
661180	16.06.2013	19:17:53	24.06.2013	10:57:33
625137	13.06.2013	18:14:29	20.06.2013	23:53:26
606864	16.06.2013	13:39:29	23.06.2013	14:13:53
576471	03.06.2013	19:49:15	10.06.2013	11:57:06
565786	05.06.2013	08:07:58	11.06.2013	21:17:44
563884	23.06.2013	06:14:17	29.06.2013	18:52:21
531099	08.06.2013	16:19:05	14.06.2013	19:50:44
528411	16.06.2013	19:01:24	22.06.2013	21:48:15
521552	22.06.2013	18:13:29	28.06.2013	19:06:01
504367	08.06.2013	22:17:05	14.06.2013	18:23:12
451931	09.06.2013	11:45:18	14.06.2013	17:17:29
394499	15.06.2013	21:27:47	20.06.2013	11:02:46
354318	05.06.2013	12:41:59	09.06.2013	15:07:17
340149	03.06.2013	22:53:28	07.06.2013	21:22:37
328384	02.06.2013	20:16:37	06.06.2013	15:29:41
315372	09.06.2013	12:39:46	13.06.2013	04:15:58
302588	05.06.2013	14:12:10	09.06.2013	02:15:18
295345	05.06.2013	07:01:37	08.06.2013	17:04:02

Figure 16: Checking for outliers

- Trip duration 3876479 on column Q: There is one individual who rented a bike from 04.06.13 to 19.07.13. That's 46 days.
- The second longest rented bike was from 02.06.13 to 04.07.13. That's 33 days.
- That's a long way from the average rented bike duration of 547 seconds (9 minutes).

Using the IQR method:

=E2/ROWS(\$A:\$A)			
C	D	E	F
Trip Duration		Number of outliers	Percentage of outliers
Q1	514	24893	2.37 %
Q3	1398		
IQR	884		
Fence Multiplier	1.5		
Inner Fence (lower)	-812		
Outer Fence (upper)	2724		

Figure 17: Detecting outliers using IQR

Figure 17 shows:

- The IQR method detected 24893 outliers in column “trip duration”.
- The number of outliers is just 2.37 % of the whole data in the column.

Using the z-score method:

=J2/ROWS(\$A:\$A)			
H	I	J	K
Trip Duration		Number of outliers	Percentage of outliers
Mean	1372.57	1700	0.16 %
STD	9028.055		
Fence Multiplier	3		
Inner Fence (lower)	-25711.6		
Outer Fence (upper)	28456.74		

Figure 18: Detecting outliers using z-score

Figure 18 shows:

- The z-score method detected 1700 outliers in column “trip duration”.
- The number of outliers is just 0.16 % of the whole data in the column.

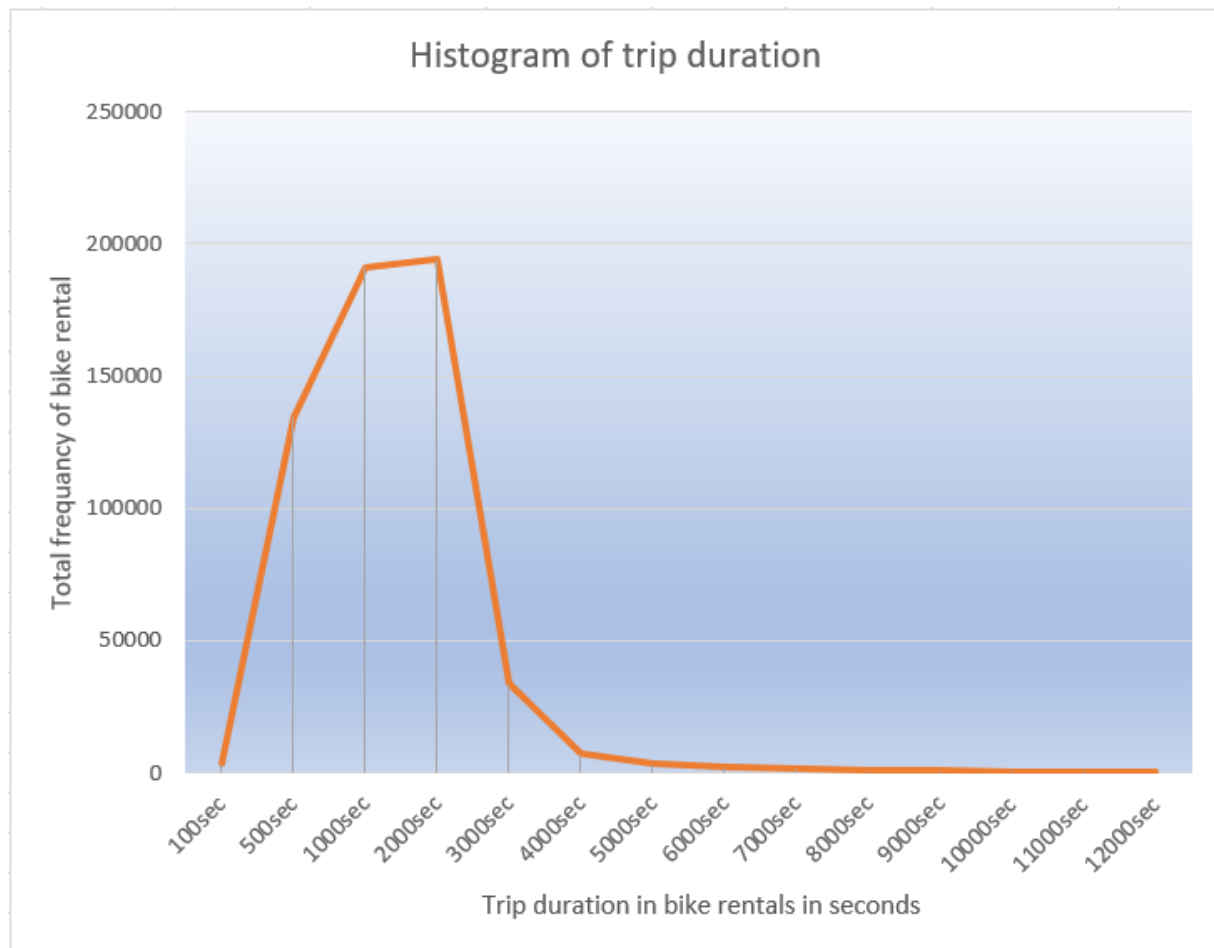


Figure 19: Histogram

The line chart shows:

- People who rents bikes usually rents it between a duration of 100 seconds (1,5 minute) to 3000 seconds (50 minutes). This is the most used duration time out of 577703 times a bike was used/rented between 01.06.13 to 01.07.13.
- Histogram shows a right skewness
- The mean is 1372.57 and higher than median 874
- As the line shows, mean better represents the central tendency of the distribution of the chart.
- Using =percentile.inc like the figure below to show if the histogram matches the percentile.

Using this histogram was based on the data in figure 20 (next page), using percentile and finding out where the trip duration was most common.



fx		=PERCENTILE.INC(\$A:\$A;D84)			
	C	D	E	F	
	25th percentile	0.25	514	This means that 25 % of the trip duration are from 514 seconds or less	
	50th percentile	0.5	874	This means that 50 % of the trip duration are from 874 seconds or less	
	75th percentile	0.75	1398	This means that 75 % of the trip duration are from 1398 seconds or less	
	95th percentile	0.95	2544	This means that 95 % of the trip duration are from 2544 seconds or less	
			25 %		
			50 %		
			75 %		
			95 %		

Figure 20: Percentile of 25<sup>th</sup>, 50<sup>th</sup>, 75<sup>th</sup> and 95<sup>th</sup>.

fx		=PERCENTRANK.INC(\$A:\$A;E84)			
	C	D	E	F	
	25th percentile	0.25	514	This means that 25 % of the trip duration are from 514 seconds or less	
	50th percentile	0.5	874	This means that 50 % of the trip duration are from 874 seconds or less	
	75th percentile	0.75	1398	This means that 75 % of the trip duration are from 1398 seconds or less	
	95th percentile	0.95	2544	This means that 95 % of the trip duration are from 2544 seconds or less	
			25 %		
			50 %		
			75 %		
			95 %		

Figure 21: Using =percentrank.inc to check if the formula and the value matches the %.

The most frequent numbers in column Trip Duration is shown below:

I used some formulas that didn't made the result I wanted, so I had to google it and used the formula in this site: <https://www.exceldemy.com/how-to-find-most-frequent-numbers-excel/>. Before this I used formulas such as =mode.sngl, =isnumber, =max, =vlookup etc. After finding the values that I was looking for, I started making a chart to show the result.

{=MODE(IF(ISERROR(MATCH(\$A\$2:\$A\$577704;\$F\$1:F1;0));\$A\$2:\$A\$577704))}						
	E	F	G	H	I	
		Top 5 numbers	Frequency		Percentile	Tripdu
		547	532		5th	
70279		484	517		25th	
97129		530	512		50th	
874		452	507		75th	
547		451	505		95th	

Figure 20: Finding the most frequent trip durations on rented bikes.

Using the top 10 most frequent numbers in a chart:

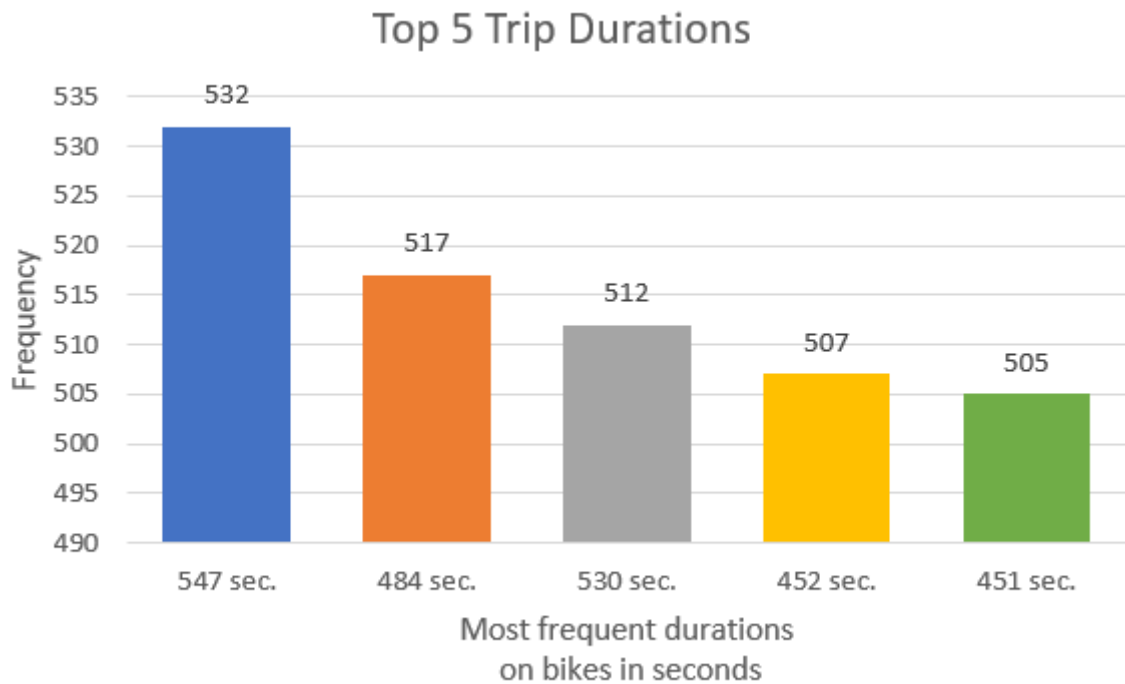


Figure 21: Top 5 trip durations

The chart shows:

- That using a rented bike for 547 seconds was the most common time a rental lasted.
- 532 people had the same amount of time spent renting a bike.

To understand this better, I could try to rather use minutes instead of seconds as shown below:

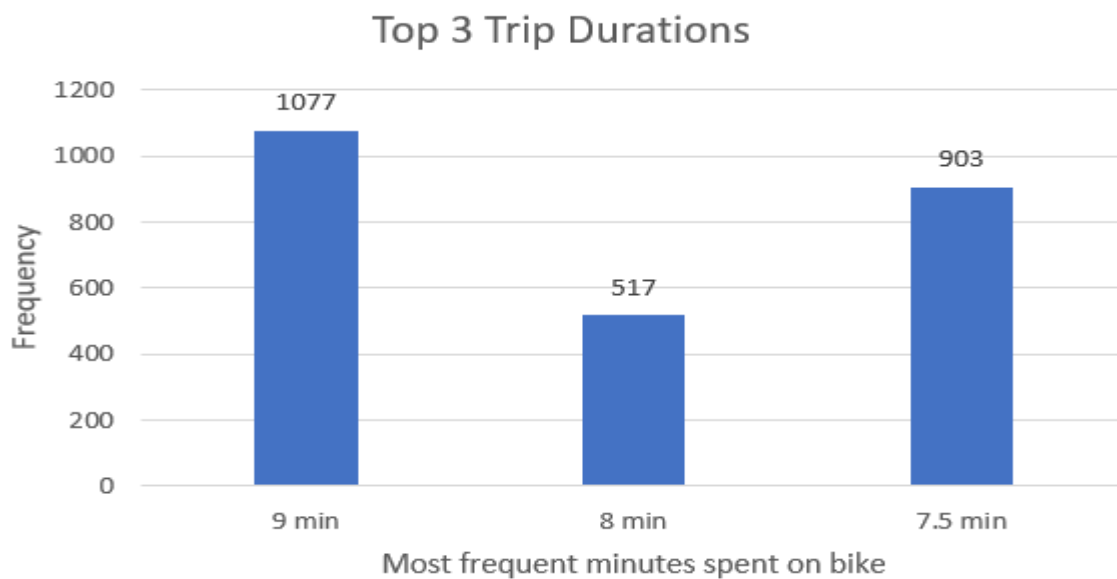


Figure 22: Simpler chart of most frequent minutes spent renting bikes

Note:

- The 9-minute bar is a combine value of the 1st and 3rd frequent duration in figure 21. The total is wrong and it should be 1044 total, not 1077.
- The 8-minute bar is the second bar in figure 21.
- The 7.5 minute bar is a combine value of the 2nd , 4th and 5th bar in figure 21.

#### Assignment 4:

Finding the Euclidean distance between column “start stations” to “end station”. This calculates the shortest possible distance between two locations.

- I made a new column called Euclidean distance:

=SQRT((A2-C2)^2+(B2-D2)^2)					
	A	B	C	D	E
	start station latitude	start station longitude	end station latitude	end station longitude	Euclidean Distance
	40.7423543	-73.98915076	40.74317449	-74.00366443	0.014536827
	40.7423543	-73.98915076	40.74317449	-74.00366443	0.014536827
	40.69512845	-73.99595065	40.69512845	-73.99595065	0
	40.73524276	-73.98758561	40.6917823	-73.9737299	0.045615702
	40.70569254	-74.01677685	40.68926942	-73.98912867	0.032158059
	40.75038009	-73.98338988	40.69512845	-73.99595065	0.056661421
	40.73454567	-73.99074142	40.710451	-73.960876	0.038373121
	40.7454973	-74.00197139	40.75096735	-73.99444208	0.009306554
	40.72229346	-73.99147535	40.68216564	-73.95399026	0.05491242
	40.7405826	-74.00550867	40.72362738	-73.99949601	0.017989763
	40.76584941	-73.98690506	40.69383	-73.990539	0.072111032
	40.714215	-73.981346	40.710451	-73.960876	0.020813183
	40.68981035	-73.97493121	40.68223166	-73.9614583	0.015458197
	40.7462009	-73.98855723	40.75096735	-73.99444208	0.007573012
	40.744023	-73.976056	40.71413089	-73.9970468	0.036526044
	40.7149787	-74.013012	40.728145	-73.990214	0.026326797
	40.7149787	-74.013012	40.728846	-74.008591	0.014554973
	40.73827428	-73.98751968	40.73827428	-73.98751968	0
	40.71939226	-74.00247214	40.76228826	-73.98336183	0.046960311
	40.7462009	-73.98855723	40.75096735	-73.99444208	0.007573012
	40.74734825	-73.99723551	40.74734825	-73.99723551	0
	40.73401143	-74.00293877	40.73492695	-73.99200509	0.010971943
	40.76095756	-73.96724467	40.73291553	-74.00711384	0.048743268
	40.76669671	-73.99061728	40.72621788	-73.98379855	0.041049126
	40.74475148	-73.99915362	40.74475148	-73.99915362	0
	40.73781509	-73.99994661	40.73649403	-73.99704374	0.003189334
	40.74394314	-73.97966069	40.73587678	-73.98205027	0.008412863
	40.72210379	-73.99724901	40.72019576	-73.98997825	0.007516949
	40.722055	-73.989111	40.72066442	-73.98517977	0.004169926
	40.74734825	-73.99723551	40.751873	-73.977706	0.020046823
	40.73935542	-73.99931783	40.76669671	-73.99061728	0.028692259
	40.7262807	-73.98978041	40.72453734	-73.98185424	0.008115632
	40.71893904	-73.99266288	40.71406667	-73.99293911	0.004880194
	40.71893904	-73.99266288	40.71893904	-73.99266288	0
	40.73291553	-74.00711384	40.73291553	-74.00711384	0
	40.71602118	-73.99974372	40.72362738	-73.99949601	0.007610232

Figure 23: Finding the Euclidean distance using formula =SQRT.

NOTE: I had to make a new sheet in fear of my excel crashing again. If not, I would have made a new column inside the sheet 201306-citibike-tripdata.

NOTE: When calculating the Euclidean and Manhattan distance I noticed that there where data that could affect assignment 5.

201306

A	B	C	D	E	F
40.69475701	-73.99052739	40.72938685	-73.97772429	0.036920796	0.04743294
40.71911552	-74.00666661	40.7262807	-73.98978041	0.018343488	0.02405138
40.73221853	-73.98165557	40.75455731	-73.96592976	0.027318898	0.03806459
40.71260486	-73.96264403	40.722055	-73.989111	0.028103481	0.03591711
40.7262807	-73.98978041	40.73221853	-73.98165557	0.010063342	0.01406267
40.69089272	-73.99612349	40.69165183	-73.9999786	0.003929137	0.00461422
40.69239502	-73.99337909	40.68382604	-73.97632328	0.01908738	0.02562479
40.69089272	-73.99612349	40.69165183	-73.9999786	0.003929137	0.00461422
40.756014	-73.967416	40.735238	-74.000271	0.038872782	0.053631
40.7284186	-73.98713956	40.74780373	-73.9734419	0.023736241	0.03308279
40.73935542	-73.99931783	0	0	84.47244592	114.7386733
40.72362738	-73.99949601	40.76370739	-73.9851615	0.042566247	0.05441452
40.75513557	-73.98658032	40.73971301	-73.99456405	0.017366499	0.02340629
40.75513557	-73.98658032	40.74195138	-74.00803013	0.025177713	0.034634
40.73261787	-73.99158043	40.70569254	-74.01677685	0.036875913	0.05212175
40.69196035	-73.96536851	40.69363137	-73.96223558	0.003550712	0.00480395
40.759107	-73.959223	40.7462009	-73.98855723	0.032047847	0.04224033
40.7643971	-73.97371465	40.73587678	-73.98205027	0.029713485	0.03685594
40.76341379	-73.99667444	40.76269882	-73.99301222	0.003731359	0.00437719
40.71602118	-73.99974372	40.71602118	-73.99974372	0	0
40.71406667	-73.99293911	40.71625008	-74.0091059	0.016313564	0.0183502
40.71406667	-73.99293911	40.73971301	-73.99456405	0.025697766	0.02727128
40.68764484	-73.96968902	40.68851534	-73.9647628	0.005002541	0.00579672
40.75019995	-73.99093085	40.746745	-74.007756	0.017176215	0.0202801
40.72066442	-73.98517977	40.71602118	-73.99974372	0.015286213	0.01920719
40.71260486	-73.96264403	40.71260486	-73.96264403	0	0
40.72362738	-73.99949601	40.72679454	-73.99695094	0.004063039	0.00571223
40.74290902	-73.97706058	40.73587678	-73.98205027	0.00862261	0.01202193
40.74290902	-73.97706058	0	0	84.45466315	114.7199696
40.7172274	-73.98802084	40.70862144	-74.00722156	0.021041155	0.02780668
40.72405549	-74.00965965	40.71602118	-73.99974372	0.012762281	0.01795024
40.71260486	-73.96264403	40.73291553	-74.00711384	0.048888519	0.06478048
40.69308257	-73.97178913	40.71117416	-74.00016545	0.033652952	0.04646791
40.71625008	-74.0091059	40.73543934	-73.99453948	0.024091664	0.03375568
40.74395411	-73.99144871	40.76009437	-73.99461843	0.01644856	0.01930998

Figure 24: Missing values 1.

The red line shows missing data from column C and D. That's missing data of "end station latitude" and "end station longitude". Without that data, the Euclidean and Manhattan distance values are useless. Because it doesn't show the true value of the distance.

Like in the yellow line, the start and end distance are identical. No wonder the Euclidean and Manhattan distances are at value 0, and the mode tells me it's the most frequent value in the data.

Let's look at some sheets including/adding some important columns so we can have more insights of the data. I'll go back to sheet 201306-citibike-tripdata.

	H	I	J	K	L	Q	R	T	U
1	end station id	end station name	end station latitude	end station longitude	bikeid	Starttime Date	Starttime Hour	Stoptime Date	Stoptime Hour
58	Unknown	Unknown	Unknown	Unknown	17470	01.06.2013	00:11:04	01.06.2013	00:20:11
104	Unknown	Unknown	Unknown	Unknown	15090	01.06.2013	00:40:27	01.06.2013	00:49:24
122	Unknown	Unknown	Unknown	Unknown	16257	01.06.2013	00:47:51	01.06.2013	00:55:43
213	Unknown	Unknown	Unknown	Unknown	20106	01.06.2013	01:32:55	01.06.2013	01:35:28
291	Unknown	Unknown	Unknown	Unknown	18792	01.06.2013	02:28:10	01.06.2013	02:42:11
304	Unknown	Unknown	Unknown	Unknown	15752	01.06.2013	02:40:31	01.06.2013	02:55:35
317	Unknown	Unknown	Unknown	Unknown	17106	01.06.2013	02:54:35	01.06.2013	03:06:59
336	Unknown	Unknown	Unknown	Unknown	16461	01.06.2013	03:22:48	01.06.2013	03:38:02
356	Unknown	Unknown	Unknown	Unknown	17283	01.06.2013	03:52:23	01.06.2013	04:04:00
375	Unknown	Unknown	Unknown	Unknown	18327	01.06.2013	04:50:53	01.06.2013	05:01:13
451	Unknown	Unknown	Unknown	Unknown	19912	01.06.2013	07:21:49	01.06.2013	11:14:14
497	Unknown	Unknown	Unknown	Unknown	18188	01.06.2013	07:50:57	01.06.2013	08:05:36
500	Unknown	Unknown	Unknown	Unknown	17548	01.06.2013	07:51:49	01.06.2013	08:06:09
513	Unknown	Unknown	Unknown	Unknown	19315	01.06.2013	07:55:40	01.06.2013	08:18:44
522	Unknown	Unknown	Unknown	Unknown	15555	01.06.2013	07:59:35	01.06.2013	17:42:21
525	Unknown	Unknown	Unknown	Unknown	18257	01.06.2013	08:00:35	01.06.2013	08:05:46
539	Unknown	Unknown	Unknown	Unknown	16781	01.06.2013	08:08:05	01.06.2013	12:13:15
616	Unknown	Unknown	Unknown	Unknown	15760	01.06.2013	08:33:05	01.06.2013	08:37:33
630	Unknown	Unknown	Unknown	Unknown	15353	01.06.2013	08:38:57	01.06.2013	08:45:33
658	Unknown	Unknown	Unknown	Unknown	15721	01.06.2013	08:48:24	01.06.2013	09:02:36

Figure 25: Missing values 2.

The figure shows:

- That even though column “end station ID”, “end station name”, “end station latitude, and “end station longitude” are missing, the data still shows values in column “trip duration”, “start date”, “stop time date” etc., as shown in figure 25.
- I counted 18059 missing rows in column H, I, J and K.
- The missing data in columns H, I, J and K makes it difficult to calculate the distance. Data is useless when calculating the distance since there is no data that tells us where in New York the bike trip ended. We have only the date and time that shows us when the bike rent ended.

Let’s look at the gray line in figure 25 as an example:

- The person renting the bike from 07:59:35 in the morning till 17:42:21 in the afternoon/evening in 01.06.13, had a trip duration of 34966 sec. That’s over 9 hours bike rent. Still, we don’t know the distance of the trip.
- Could be that the trip didn’t ended in an end station, but some random place in the city.

I think that the missing values in distance should not be in the data in finding the values of the distance in assignment 5. The 18059 rows will not be included. Because as shown in figure 23 and 24, the data of distance will be misleading.

If we look again in figure 24, the yellow lines are also values that should not be in the data in finding the distance, since the person who rented the bike never left the start station. I’ll show some examples under:

	A	D	E	F	G	H	I	J	K	Q	R	T	U
ripduration	start station id	start station name	start station latitude	start station longitude	end station id	end station name	end station latitude	end station longitude	Starttime Date	Starttime Hour	Stoptime Date	Stoptime Hour	
695	444	Broadway & W 24 St	40.7423543	-73.98915076	434	9 Ave & W 18 St	40.74317449	-74.00366443	01.06.2013	00:00:01	01.06.2013	00:11:36	
693	444	Broadway & W 24 St	40.7423543	-73.98915076	434	9 Ave & W 18 St	40.74317449	-74.00366443	01.06.2013	00:00:08	01.06.2013	00:11:41	
2059	406	Hicks St & Montague St	40.69512845	-73.99595065	406	Hicks St & Montague St	40.69512845	-73.99595065	01.06.2013	00:00:44	01.06.2013	00:35:03	
123	475	E 15 St & Irving Pl	40.73524276	-73.98758561	262	Washington Park	40.6917823	-73.9737299	01.06.2013	00:01:04	01.06.2013	00:03:07	
1521	2008	Little West St & 1 Pl	40.70569254	-74.01677685	310	State St & Smith St	40.68926942	-73.98912867	01.06.2013	00:01:22	01.06.2013	00:26:43	
2028	485	W 37 St & 5 Ave	40.75038009	-73.98338988	406	Hicks St & Montague St	40.69512845	-73.99595065	01.06.2013	00:01:47	01.06.2013	00:35:35	
2057	285	Broadway & E 14 St	40.73454567	-73.99074142	532	S 5 Pl & S 5 St	40.710451	-73.960876	01.06.2013	00:02:33	01.06.2013	00:36:50	
369	509	9 Ave & W 22 St	40.7454973	-74.00197139	521	8 Ave & W 31 St N	40.75096735	-73.99444208	01.06.2013	00:03:29	01.06.2013	00:09:38	
1829	265	Stanton St & Chrystie St	40.72229346	-73.99147535	436	Hancock St & Bedford Ave	40.68216564	-73.95399026	01.06.2013	00:03:47	01.06.2013	00:34:16	
829	404	9 Ave & W 14 St	40.7405826	-74.00550867	303	Mercer St & Spring St	40.72362738	-73.99949601	01.06.2013	00:04:22	01.06.2013	00:18:11	
1316	423	W 54 St & 9 Ave	40.76584941	-73.98690506	314	Cadman Plaza West & Montague St	40.69383	-73.990539	01.06.2013	00:04:28	01.06.2013	00:26:24	
1456	502	Henry St & Grand St	40.714215	-73.981346	532	S 5 Pl & S 5 St	40.710451	-73.960876	01.06.2013	00:04:41	01.06.2013	00:28:57	
386	241	DeKalb Ave & S Portland Ave	40.68981035	-73.97493121	365	Fulton St & Grand Ave	40.68223166	-73.9614583	01.06.2013	00:05:13	01.06.2013	00:11:39	
924	486	Broadway & W 29 St	40.7462009	-73.98855723	521	8 Ave & W 31 St N	40.75096735	-73.99444208	01.06.2013	00:05:21	01.06.2013	00:20:45	
1233	527	E 33 St & 2 Ave	40.744023	-73.976056	296	Division St & Bowery	40.71413089	-73.9970468	01.06.2013	00:06:44	01.06.2013	00:27:17	
512	309	Murray St & West St	40.7149787	-74.013012	300	Shevchenko Pl & E 7 St	40.728145	-73.990214	01.06.2013	00:06:33	01.06.2013	00:12:08	
505	309	Murray St & West St	40.7149787	-74.013012	347	Greenwich St & W Houston St	40.728846	-74.008591	01.06.2013	00:03:45	01.06.2013	00:12:10	
833	503	E 20 St & Park Ave	40.73827428	-73.98751968	503	E 20 St & Park Ave	40.73827428	-73.98751968	01.06.2013	00:07:29	01.06.2013	00:21:22	
1818	257	Lispenard St & Broadway	40.71939226	-74.00247214	500	Broadway & W 51 St	40.76228826	-73.98336183	01.06.2013	00:08:10	01.06.2013	00:38:28	
682	486	Broadway & W 29 St	40.7462009	-73.98855723	521	8 Ave & W 31 St N	40.75096735	-73.99444208	01.06.2013	00:08:53	01.06.2013	00:20:15	
899	494	W 26 St & 8 Ave	40.74734825	-73.99723551	494	W 26 St & 8 Ave	40.74734825	-73.99723551	01.06.2013	00:09:25	01.06.2013	00:24:24	
626	380	W 4 St & 7 Ave S	40.73401143	-74.00293877	382	University Pl & E 14 St	40.73492695	-73.99200509	01.06.2013	00:09:52	01.06.2013	00:20:18	
219	305	E 58 St & 3 Ave	40.76095756	-73.96724467	358	Christopher St & Greenwich St	40.73291553	-74.00711384	01.06.2013	00:10:28	01.06.2013	00:14:07	

Figure 25: Missing values 3

The figure shows:

NOTE: On figure 24 the yellow lines are now replaced with red lines. I had some issues loading excel again and had to repeat the analysis.

- The two red lines shows rented bikes with exactly the same start latitude/longitude as in end latitude/longitude.
- The upper red line in figure 25: Even though rented time lasted from 00:00:44 till 00:35:03 nighttime with a trip duration of 2059 (34 minutes), the bike never leaved the start station. The figure shows start station and end station are the same.
- I counted 23625 after getting value of 0 in calculating both Euclidean and Manhattan distance.

This mean that those data can't be in the calculation of distance since I believe the trip never left the start station. Trip duration only show the amount of time spent renting the bike.

This leaves me to not include 38443 of 577703 data observations in the citibike-tripdata in finding the different values in distance.

NOTE: I had big issues in excel when filling the blank rows. I used a lot of time finding ways to make it work. Like in trip durations, I copied and pasted the distance in a new workbook, but it kept crashing and loading to infinity...

Distance Citibike - Excel (Svarer ikke)

	A	B	C	D	E	F	G
1	Start station latitude	Start station longitude	End station latitude	End station longitude	Euclidean distance	Manhattan distance	
2	40.7423543	-73.98915076	40.74317449	-74.00366443	0.014536827	0.01533386	
3	40.7423543	-73.98915076	40.74317449	-74.00366443	0.014536827	0.01533386	
4							
5	40.73524276	-73.98758561	40.6917823	-73.9737299	0.045615702	0.05731617	
6	40.70569254	-74.01677685	40.68926942	-73.98912867	0.032158059	0.0440713	
7	40.75038009	-73.98338988	40.69512845	-73.99595065	0.056661421	0.06781241	
8	40.73454567	-73.99074142	40.710451	-73.960876	0.038373121	0.05396009	
9	40.7454973	-74.00197139	40.75096735	-73.99444208	0.009306554	0.01299936	
10	40.72229346	-73.99147535	40.68216564	-73.95399026	0.05491242	0.07761291	
11	40.7405826	-74.00550867	40.72362738	-73.99949601	0.017989763	0.02296788	
12	40.76584941	-73.98690506	40.69383	-73.990539	0.072111032	0.07565335	
13	40.714215	-73.981346	40.710451	-73.960876	0.020813183	0.024234	
14	40.68981035	-73.97493121	40.68223166	-73.9614583	0.015458197	0.0210516	
15	40.7462009	-73.98855723	40.75096735	-73.99444208	0.007573012	0.0106513	
16	40.744023	-73.976056	40.71413089	-73.9970468	0.036526044	0.05088291	
17	40.7149787	-74.013012	40.728145	-73.990214	0.026326797	0.0359643	
18	40.7149787	-74.013012	40.728846	-74.008591	0.014554973	0.0182883	
19							
20	40.71939226	-74.00247214	40.76228826	-73.98336183	0.046960311	0.06200631	
21	40.7462009	-73.98855723	40.75096735	-73.99444208	0.007573012	0.0106513	
22							
23	40.73401143	-74.00293877	40.73492695	-73.99200509	0.010971943	0.0118492	
24	40.76095756	-73.96724467	40.73291553	-74.00711384	0.048743268	0.0679112	
25	40.76669671	-73.99061728	40.72621788	-73.98379855	0.041049126	0.04729756	
26							
27	40.73781509	-73.99994661	40.73649403	-73.99704374	0.003189334	0.00422393	
28	40.74394314	-73.97966069	40.73587678	-73.98205027	0.008412863	0.01045594	

**Note:**

*This was a big issue. I spent hours in figuring how to get excel to work with me. It just kept crashing and loading. It stopped completely when filling the empty rows (figure above) after deleting the values/missing values that I believed didn't fit in finding the distance. I used all the tips in Discord, changed to new workbook, copy/paste single columns for analysis, etc. etc., but it kept crashing. I even changed computers. I used a lot of time on this. Did all the work finding the right way to analyze the distance and couldn't continue my work. I'm so sorry I didn't finish assignment 5 because of this issue.*



## Assignment 6:

### A naïve prediction:

In this assignment I made a new sheet called predictions. First prediction was to make a naïve prediction. I made three columns, Day and Rentals (as shown in the assignment pdf), then made a column called Quantity Prediction.

	C	D	E
	Day	Rentals	Quantity Prediction
1		8722	
2		15971	
3		7598	
4		15782	
5		15690	
6		12420	
7		1226	
8		18007	
9		21034	
10		3897	
11		12596	
12		16795	
13		8016	
14		14942	
15		25178	
16		21966	
17		19659	
18		13724	
19		24142	
20		26603	
21		27997	
22		29281	
23		28065	
24		24251	
25		27827	
26		29499	
27		24334	
28		27818	
29		29753	
30		24909	24909
1			24909
2			24909
3			24909
4			24909
5			24909
6			24909
7			24909

Figure 26: Making a naïve prediction of rental bikes.

- To find values inside column rentals, I used =countif
- Finding prediction on 01.07.13 to 07.07.13 I used =\$D\$31.
- Since the number of rental bikes in 30.06.13 was 24909, the naïve prediction will predict that the same amount of bikes will be rented from 01.07.13 to 07.07.13.
- The chart is shown below.

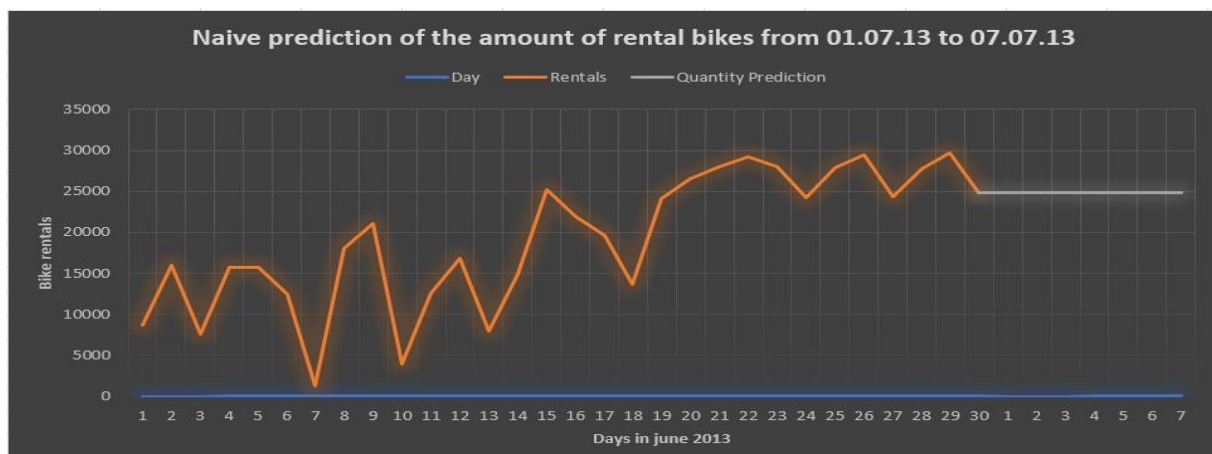




Figure 27: Naïve prediction.

The chart shows:

- That the prediction value from 01.07. to 07.07. as the same value and therefore a straight line.
- This shows a prediction but maybe not quite realistic.

### The average forecast:

This prediction focuses on the average or the mean value. Finding the mean value using =average on column D "Rentals", gives me a value of 19257. This will be the value of prediction the first 7 days in juli.

5.2.013	25	21821	
5.2.013	26	29499	
5.2.013	27	24334	
5.2.013	28	27818	
5.2.013	29	29753	
5.2.013	30	24909	24909
5.2.013	1		19256.73
5.2.013	2		19256.73
5.2.013	3		19256.73
5.2.013	4		19256.73
5.2.013	5		19256.73
5.2.013	6		19256.73
5.2.013	7		19256.73
5.2.013			
5.2.013			
5.2.013			

Figure 28: Making an average forecast of rented bikes.

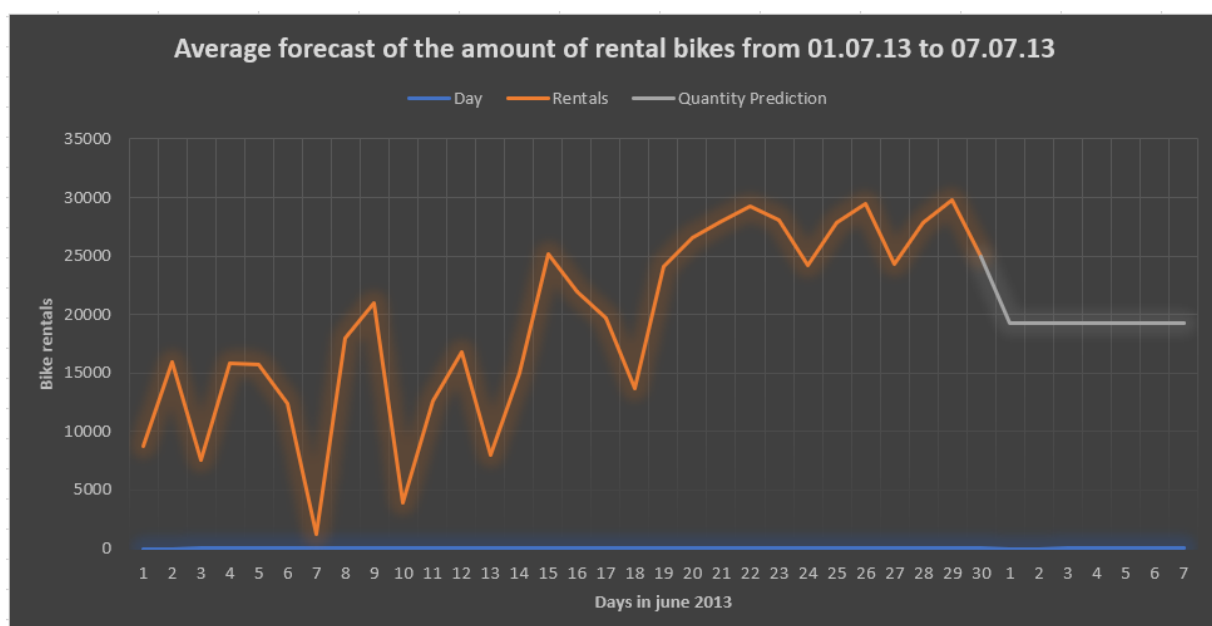


Figure 29: Average forecast.

The chart shows:

- A straight line like the naïve prediction, but with a different value.
- The same as chart figure 27, this maybe not quite realistic.

Conclusion of the two charts:

Both the naïve prediction and the average forecast on predicting forward values represents little qualitative insights. Both predictions shows us just a linear line of value even though the lines are of some difference.

### Adaptive average forecast:

In this forecast I started to find the average of the first 7 values (7 days) in column D. Then I made a new column called “adaptive average forecast using =average starting from day 8 all the way to the date 07.07.

		=AVERAGE(D2:D9)	
C	D	E	F
Day	Rentals	Adaptive average forecast	
1	8722		
2	15971		
3	7598		
4	15782		
5	15690		
6	12420		
7	1226		
8	18007	11927	
9	21034	13466	
10	3897	11956.75	
11	12596	12581.5	
12	16795	12708.125	
13	8016	11748.875	
14	14942	12064.125	
15	25178	15058.125	
16	21966	15553	
17	19659	15381.125	
18	13724	16609.5	
19	24142	18052.75	
20	26603	19278.75	
21	27997	21776.375	
22	29281	23568.75	
23	28065	23929.625	
24	24251	24215.25	
25	27827	25236.25	
26	29499	27208.125	
27	24334	27232.125	
28	27818	27384	
29	29753	27603.5	
30	24909	27057	
1		26913	
2		27356.66667	
3		27262.6	
4		26703.5	
5		27493.33333	
6		27331	
7		24909	

Figure 30: Making adaptive average forecast

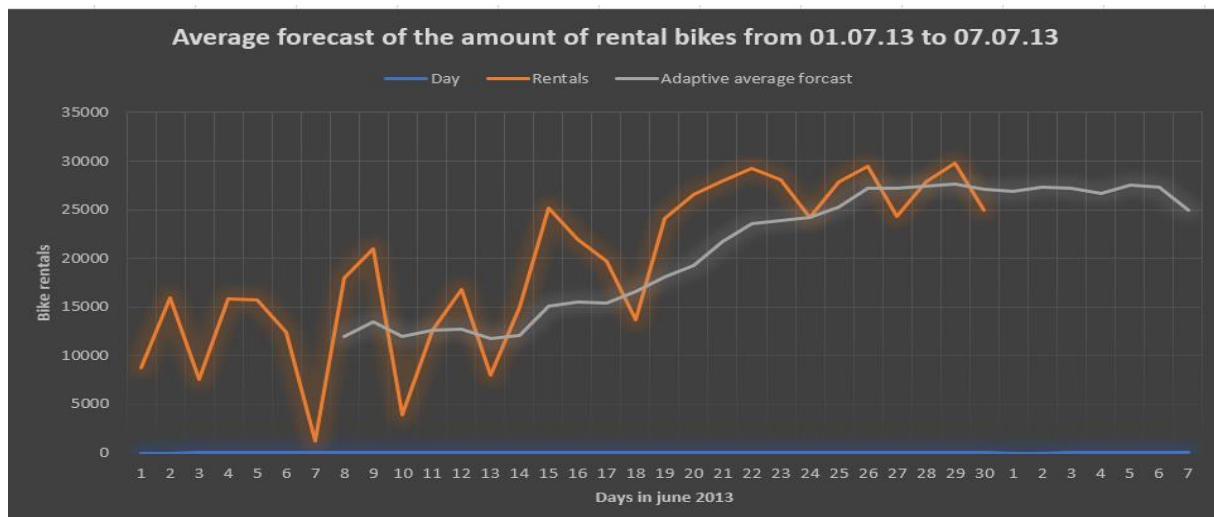


Figure 31: Adaptive mean forecast

The chart shows:

- By finding the first 7 days average value, the chart shows us two lines, where the adaptive average forecast (gray line) predicted the past, while the orange shows us what actually happened. The gray line also predicts the first 7 days in July.

### Drift projection:

This prediction is to show the trending values of past data.

✓	$f_x$	{=E31+(E31-\$D\$2)/((ROW(E31)-1)-1)}		
	C	D	E	F
	Day	Rentals	Drift projection	
1		8722		
2		15971		
3		7598		
4		15782		
5		15690		
6		12420		
7		1226		
8		18007		
9		21034		
10		3897		
11		12596		
12		16795		
13		8016		
14		14942		
15		25178		
16		21966		
17		19659		
18		13724		
19		24142		
20		26603		
21		27997		
22		29281		
23		28065		
24		24251		
25		27827		
26		29499		
27		24334		
28		27818		
29		29753		
30		24909	24909	
1			25467.17241	
2			26025.34483	
3			26583.51724	
4			27141.68966	
5			27699.86207	
6			28258.03448	
7			28816.2069	

Figure 32: Making a drift projection chart.

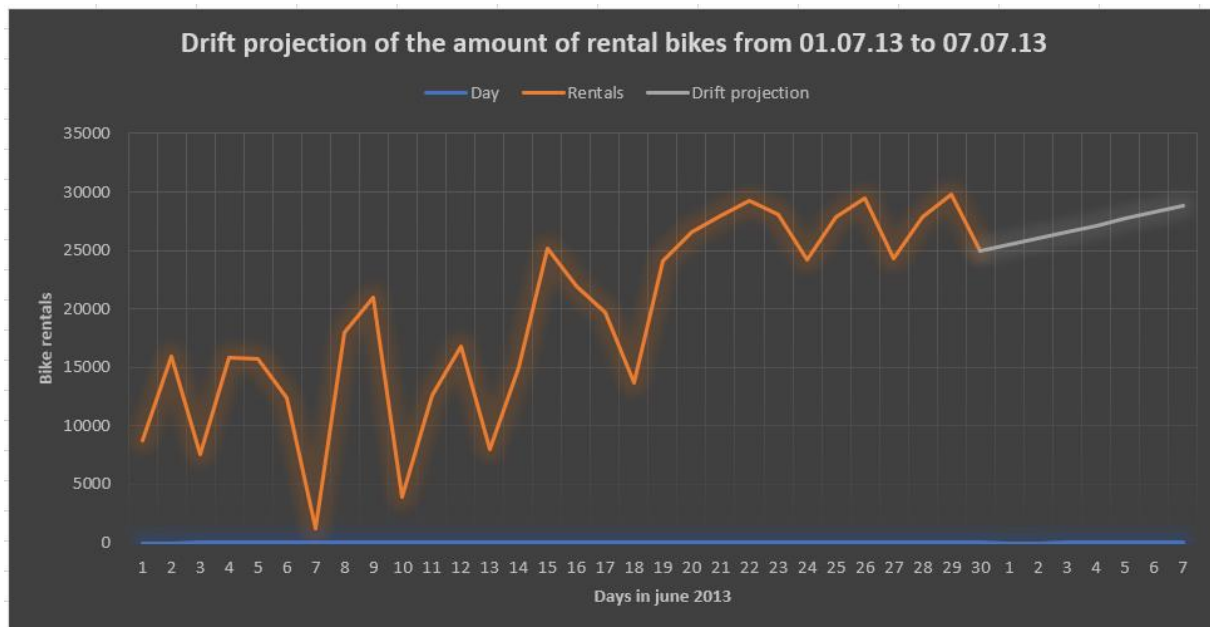


Figure 33: Drift projection chart.

The chart shows:

- An upwards line trending value.
- The trend of increased rented bikes in the past will make the drift projection go up. This is because the prediction follows the trend.

#### Naïve seasonal method:

To make the naïve seasonal method I copied the values from date 01.06 to 07.06. and pasted it from 01.07. to 07.07.

✓	<i>f<sub>x</sub></i>	=D2	
	C	D	E
	Day	Rentals	Naive sesonal predic
1		8722	
2		15971	
3		7598	
4		15782	
5		15690	
6		12420	
7		1226	
8		18007	
9		21034	
10		3897	
11		12596	
12		16795	
13		8016	
14		14942	
15		25178	
16		21966	
17		19659	
18		13724	
19		24142	
20		26603	
21		27997	
22		29281	
23		28065	
24		24251	
25		27827	
26		29499	
27		24334	
28		27818	
29		29753	
30		24909	24909
1			8722
2			15971
3			7598
4			15782
5			15690
6			12420
7			1226

Figure 34: Making a naïve seasonal chart.

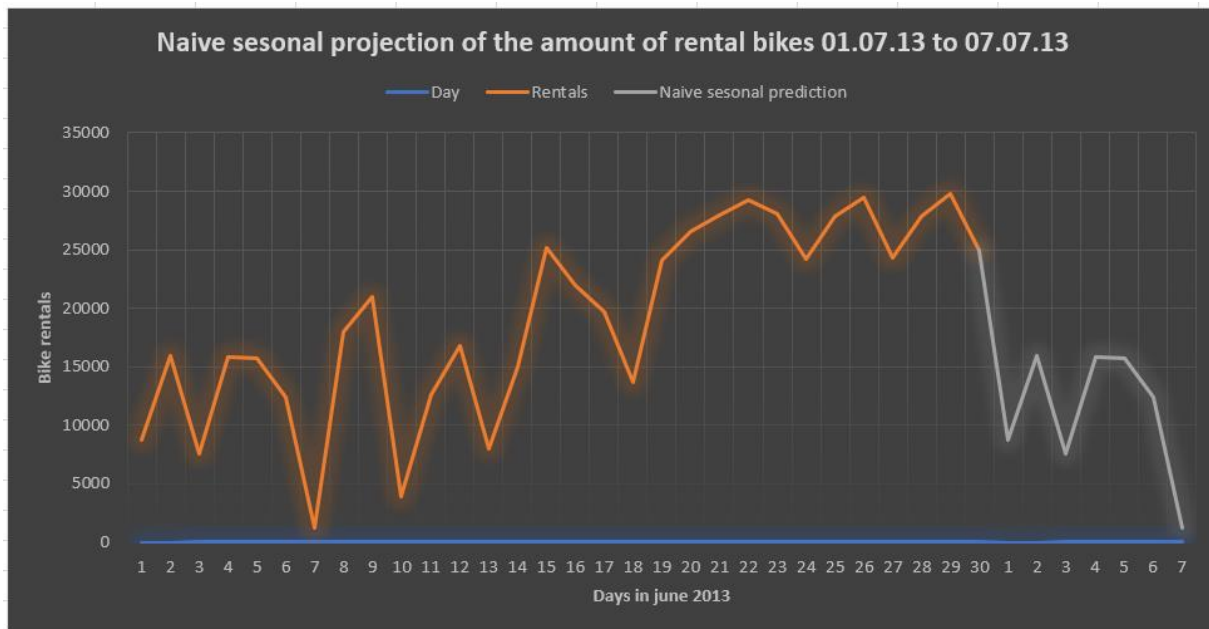


Figure 35: Naïve seasonal prediction.

The chart shows:

- The prediction shows us the same value as in the values from 01.06. to 07.06.
- The prediction shows us a falling rate of rented bikes. But if it follows the same pattern as in June, the line will go up the following day.

### Assignment 7:

Using a linear trend and exponential trend inside the charts.

Naïve seasonal projection:

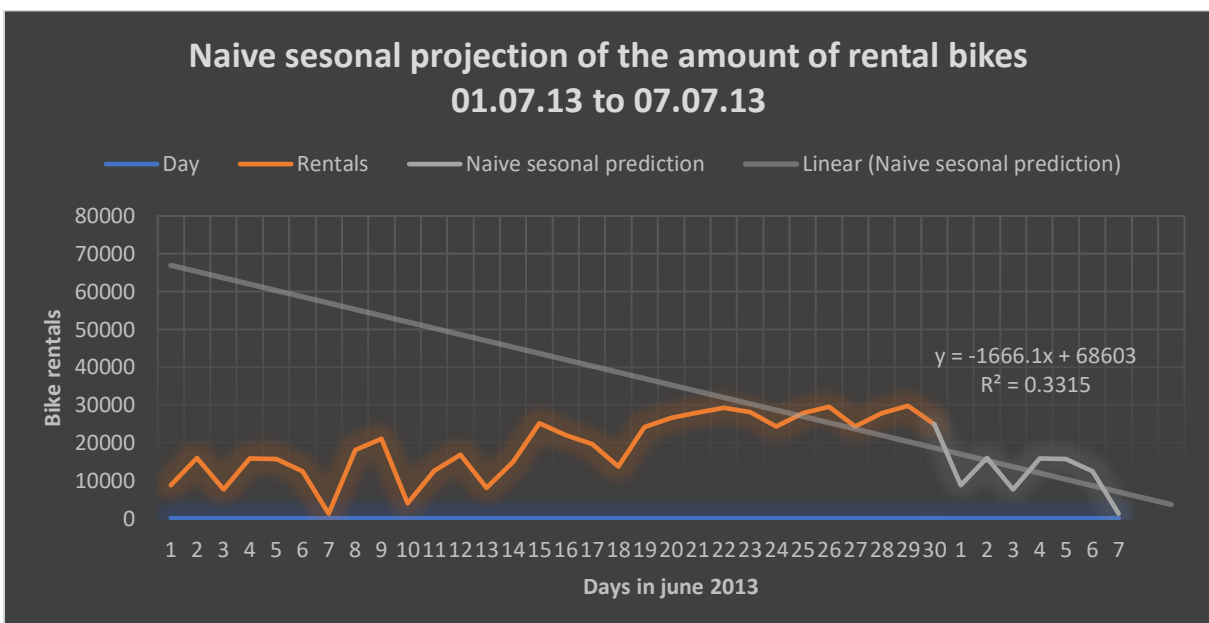


Figure 36: Linear trend telling us that the value is decreasing.

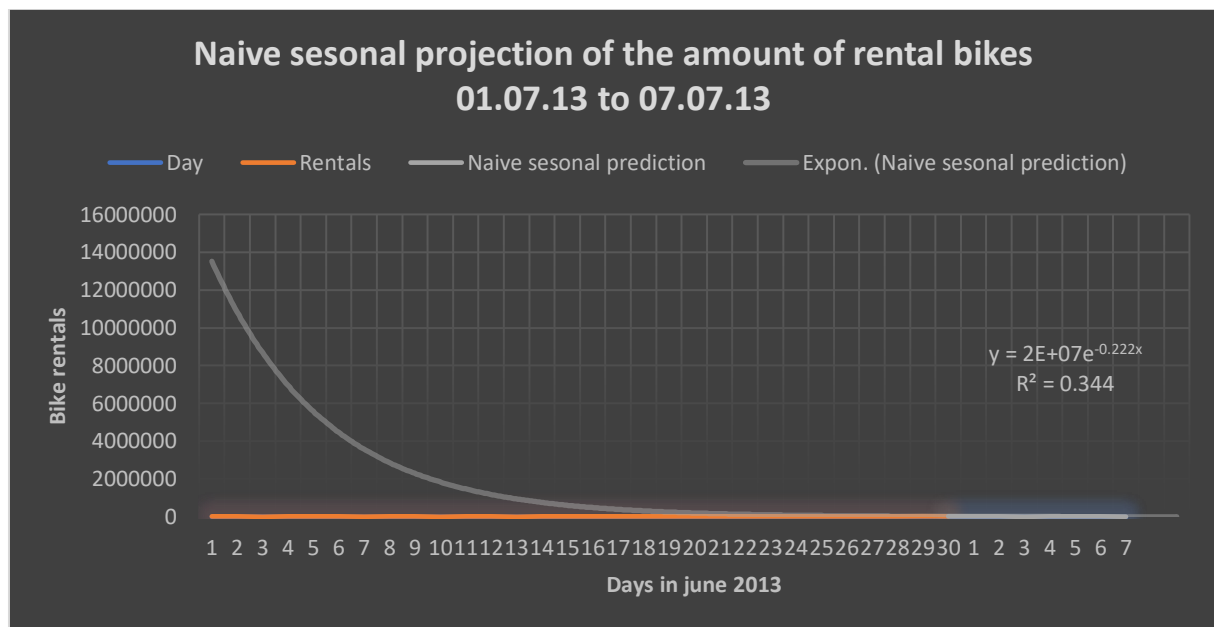


Figure 37: Exponential trend showing an increasing high rate of downfall in values. Not quite sure if I did this right....

#### Drift projection:

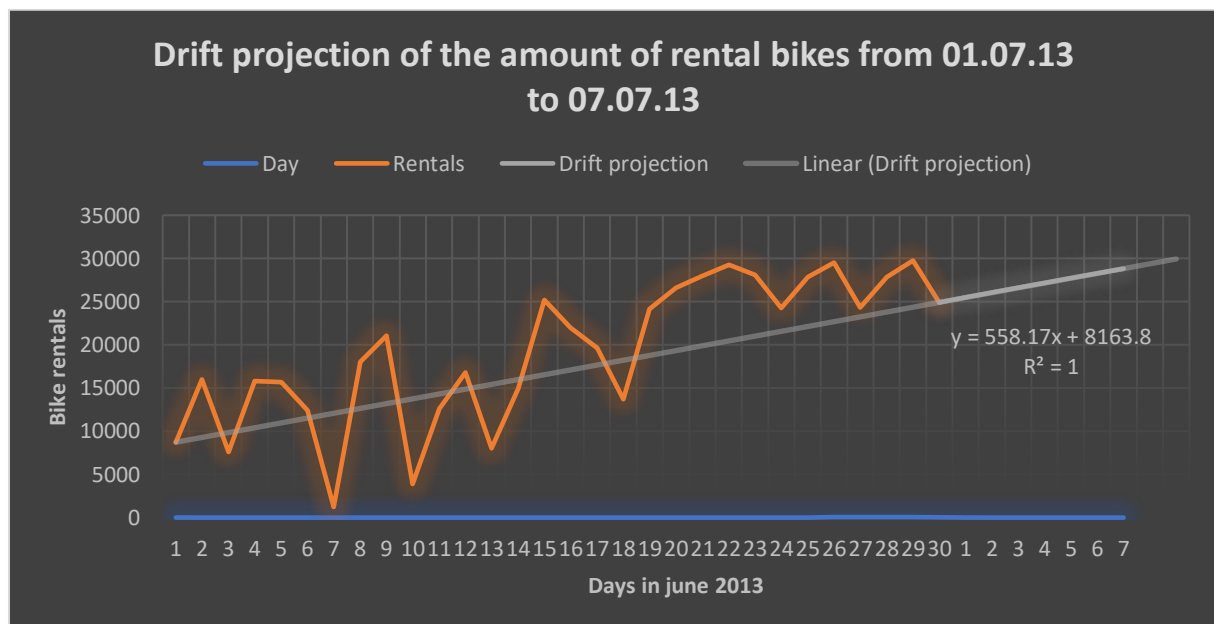


Figure 38: Linear line shows that the values are increasing with a steady rate.

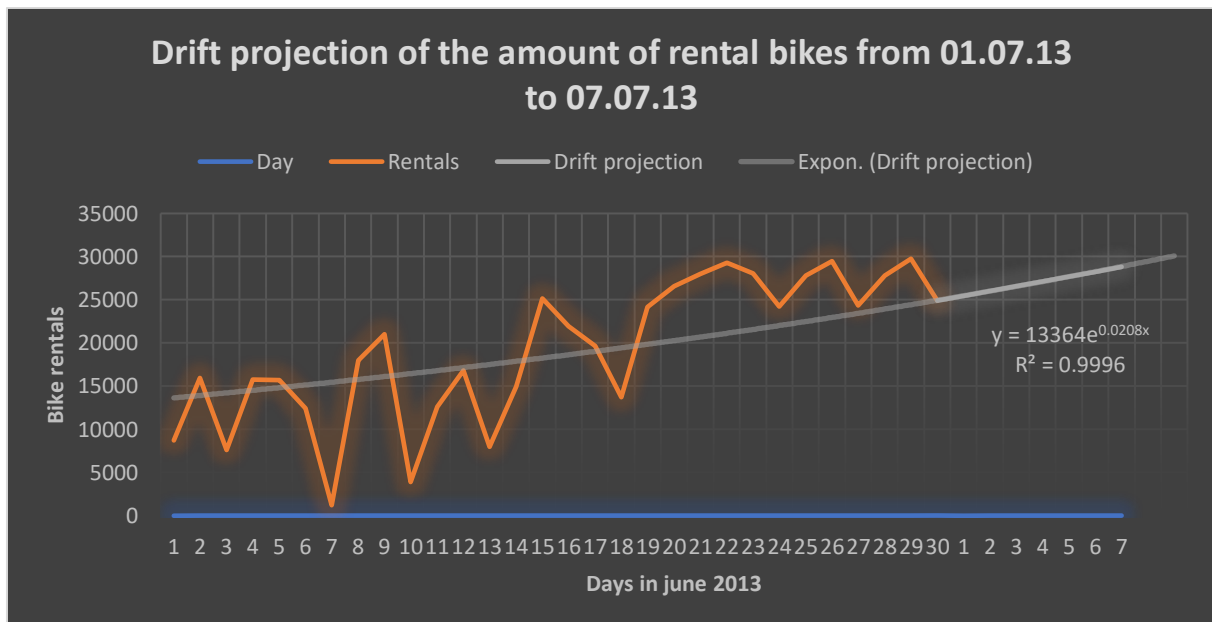


Figure 39: Exponential trend showing a rise in rates.

#### Adaptive forecast:

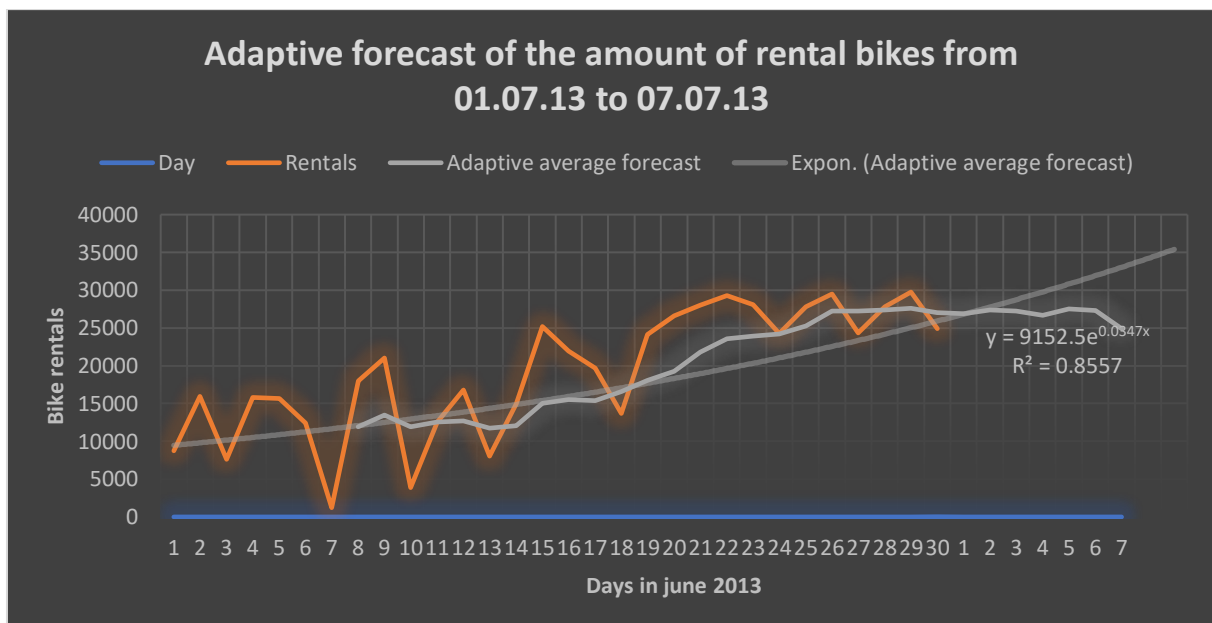


Figure 40: Exponential trend showing an increase in values.



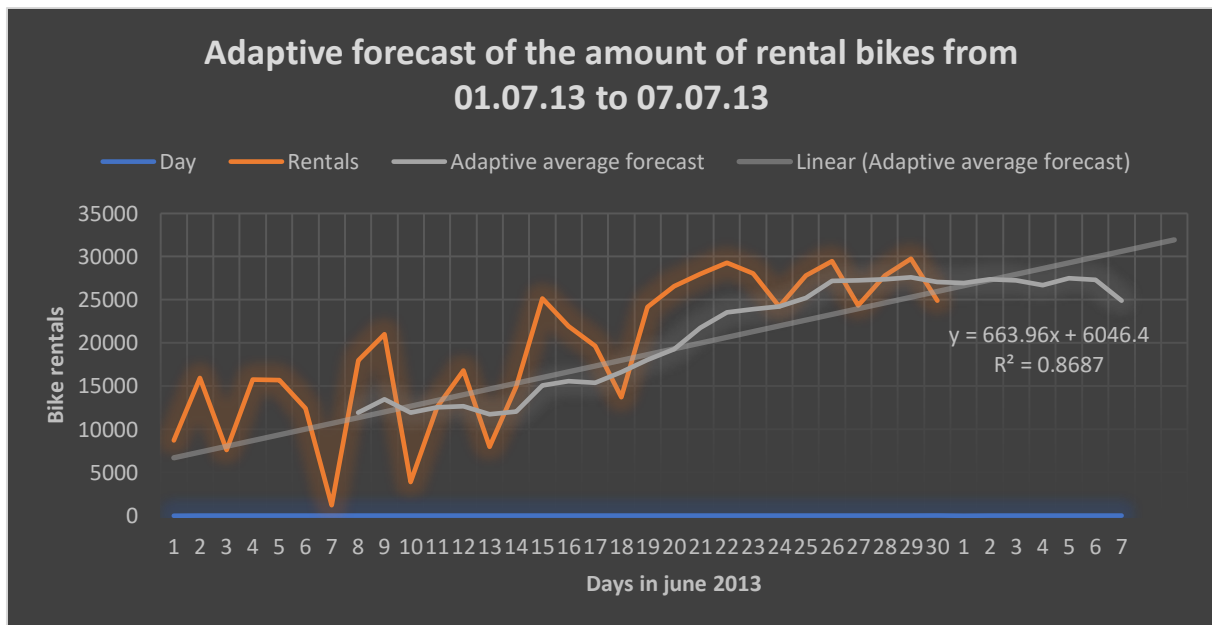


Figure 41: Like in exponential trend, the linear trend also shows a steady increase in values.

#### Average forecast:

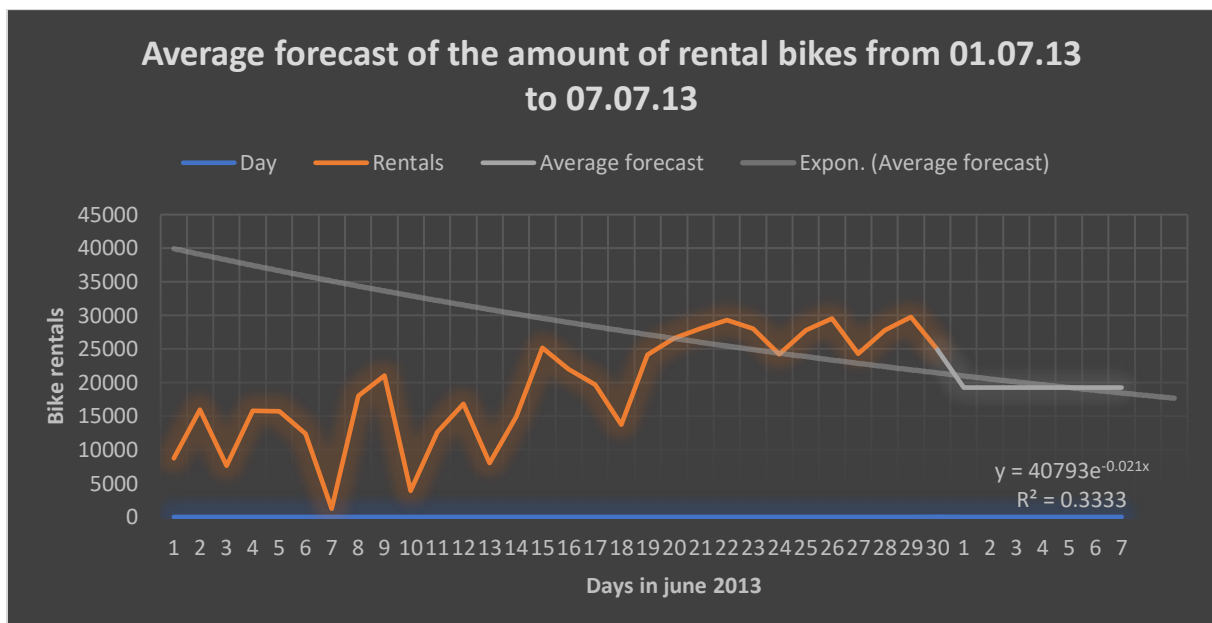


Figure 42: Exponential trend that shows a decrease in values.

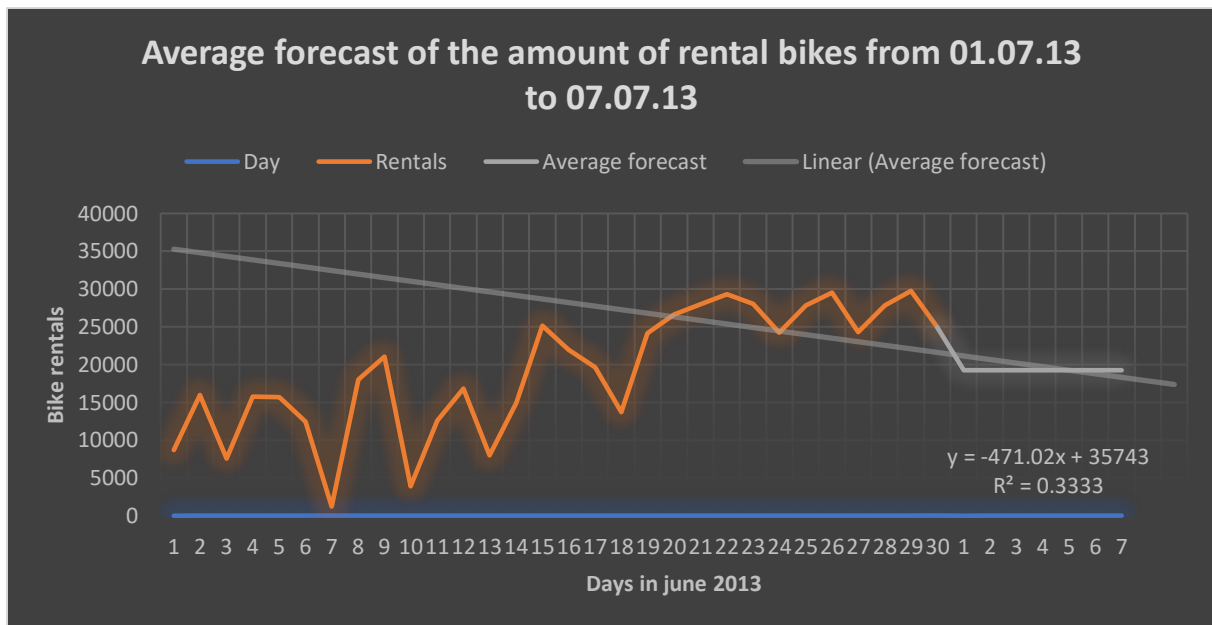


Figure 43: Like in figure 42, the linear trend shows a steady decreasing rate.

#### Naïve prediction:

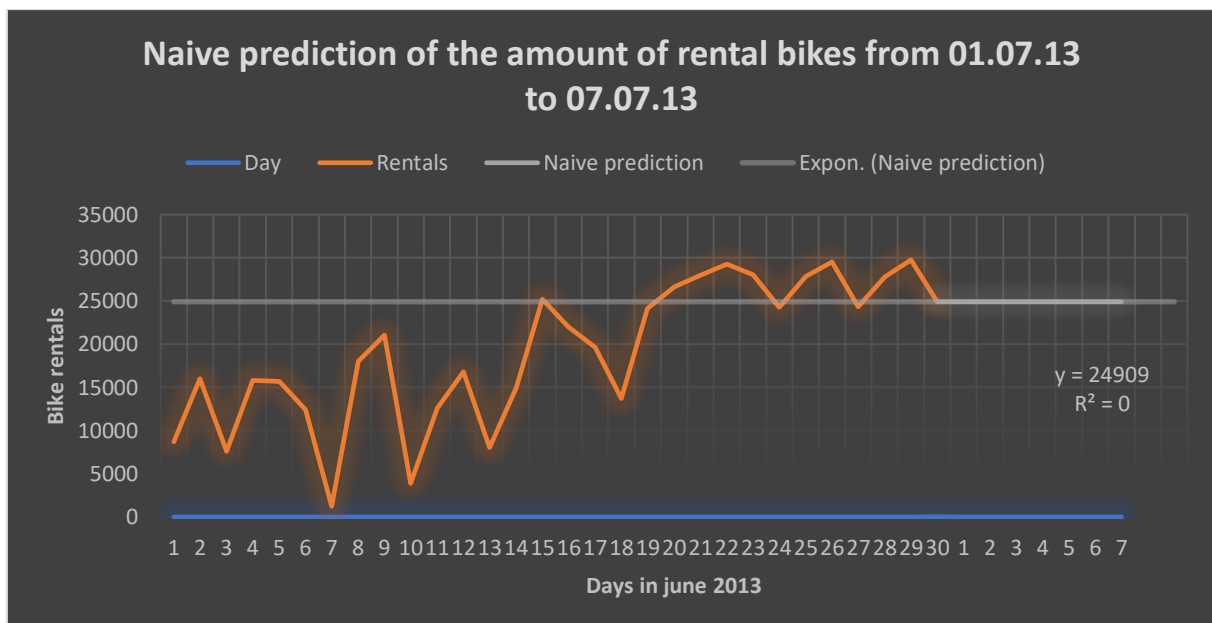


Figure 44: Exponential trend is not suitable in this chart because the naïve prediction forces the trendline to be at a zero value.

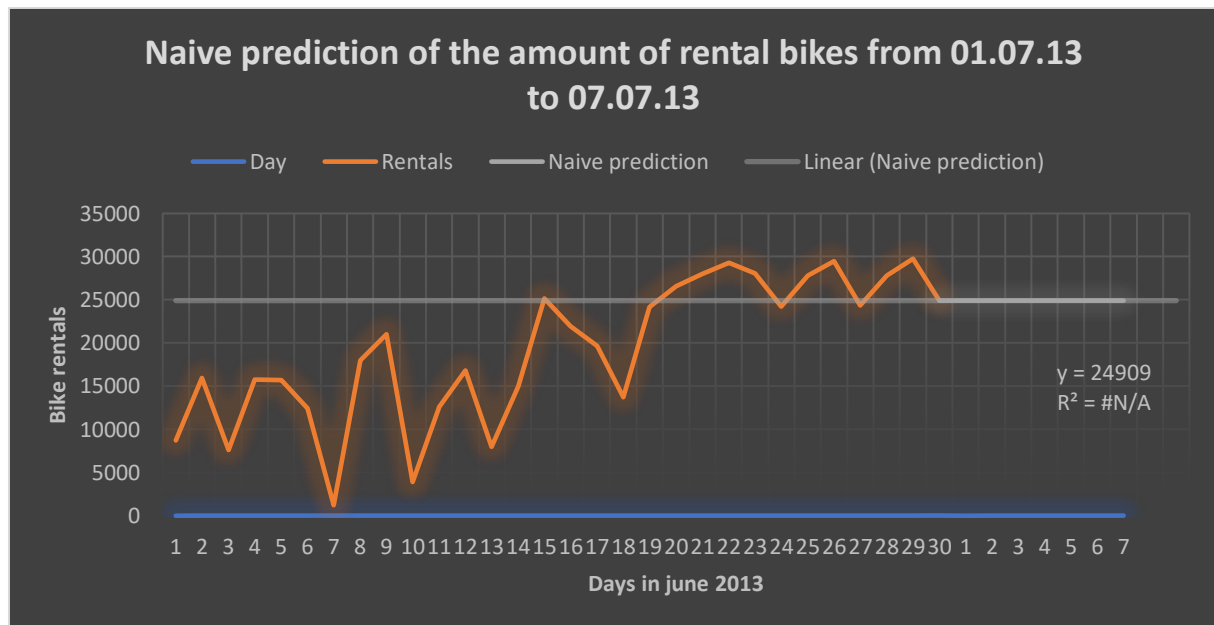


Figure 45: The tend line is constant and does not show us a decrease or increase because of the prediction.