# Big Data in NYC for Taxi rides management

with Apache Hive

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### **The NYC Taxi and Limousine Commission**

The New York City Taxi and Limousine Commission (NYC TLC) is an agency of the New York City government that licenses and regulates the medallion taxis and for-hire vehicle industries, including app-based companies. The TLC's regulatory landscape includes medallion (yellow) taxicabs, green or Boro taxicabs, black cars (including both traditional and app-based services), community-based livery cars, commuter vans, paratransit vehicles (ambulettes), and some luxury limousines [1].

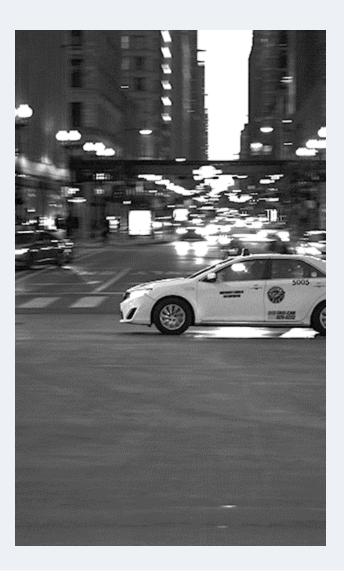
But what is the difference between yellow and green taxi (cab in American English)?

The famous NYC yellow taxis provide transportation exclusively through street-hails.

The number of taxicabs is limited by a finite number of medallions issued by the TLC.

It's possible to access this mode of transportation by standing in the street and hailing an available taxi with your hand. The pickups are not pre-arranged.

The green taxis know as well as "Boro" cab or Street Hail Livery (SHL) are permitted to accept street-hails above 110th Street in Manhattan and in the outer-boroughs of New York City. The SHL program allows livery vehicle owners to license and outfit their vehicles with green borough taxi branding, meters, credit card machines, and ultimately the right to accept street hails in addition to prearranged rides [2].



# **Requests**

Designing and implement in Hadoop MapReduce:

- 4. A HiveQL query returning the total number of passengers in yellow and green taxi in 2018.
- 5. A HiveQL query returning, for each rate code, the total amount charged to passengers of yellow and green taxi in 2018.
- 6. A HiveQL query returning, for each *PULocationID*, the list of related *DOLocationID* for yellow taxi in 2018, ordered by increasing average trip distance.

# **Data dictionary**

The two kind of file of interest, yellow and green have a different header, which means that the software has the need to distinguish and operate the two cases differently.

Moreover, the green and yellow taxis dataset produced from January 2016 to June 2016, follow a different data dictionary that requires some variation to the parsing procedure for the points 4 and 5, while for the point 6, the dataset cannot be use because of the absence of these fields.

The fields of interest to be extracted from both the yellow and green cabs files are:

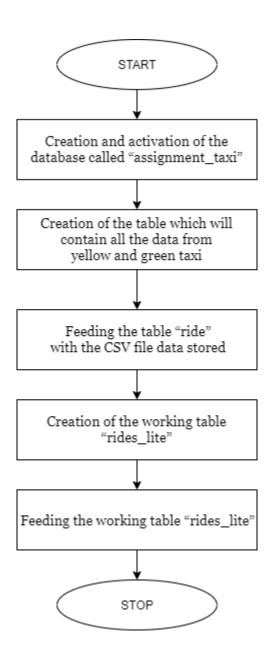
Field	Description
<pre>lpep_pickup_datetime or rpep_pickup_datetime</pre>	The pickup time, useful to get the year.
Passenger_count	The number of passengers in the vehicle.
RatecodeID	The final rate code in effect at the end of the trip.  1 = Standard rate 2 = JFK 3 = Newark 4 = Nassau or Westchester 5 = Negotiated fare 6 = Group ride
total_amount	The total amount charged to passengers. Does not include cash tips.

While the fields of interest exclusively extracted from yellow files are:

Field	Description
PULocationID	TLC Taxi Zone in which the taximeter was
	engaged.
DOLocationID	TLC Taxi Zone in which the taximeter was
	disengaged.
trip_distance	The elapsed trip distance in miles reported by
	the taximeter.

A comprehensive description of all the fields is available at nyc.gov website both for yellow [3] and the green [4] cabs.

### **DDL** and **DML** commands



### Creation and activation of the database called "assignment\_taxi"

```
CREATE DATABASE IF NOT EXISTS assignment_taxi;
USE assignment_taxi;
```

# Creation of the table which will contain all the data from yellow and green taxi

```
CREATE TABLE rides (
VendorID INT,
lpep pickup datetime STRING,
lpep dropoff datetime STRING,
store and fwd flag STRING,
RatecodeID INT,
PULocationID INT,
DOLocationID INT,
passenger count INT,
trip distance FLOAT,
fare amount INT,
extra INT,
mta tax INT,
tip amount INT,
tolls amount INT,
ehail fee STRING,
improvement surcharge INT,
total amount INT,
payment type INT,
trip type INT
) ROW FORMAT DELIMITED FIELDS TERMINATED BY ',' tblproperties
("skip.header.line.count"="2");
```

- ROW FORMAT DELIMITED FIELDS TERMINATED BY ',' sets the delimiter of the columns to be a comma.
- tblproperties ("skip.header.line.count"="2") makes the process of data importing skipping the first two rows of the CSV document, which are made up of a header and a blank row.

# Feeding the table "ride" with the CSV file data stored

```
LOAD DATA LOCAL INPATH '/media/sf_condiv/Hive_7Dec/input' OVERWRITE INTO TABLE rides;
```

### *Creation of the working table "rides\_lite"*

To make the queries faster and reducing the memory usage at the same time, a smaller version of the loaded table is created, containing only the fields useful for the queries (the ones described in <u>Data</u> dictionary section).

```
CREATE TABLE rides_lite (
year INT,
RatecodeID INT,
PULocationID INT,
DOLocationID INT,
passenger_count INT,
trip_distance FLOAT,
total_amount INT,
taxi_type INT
);
```

• The taxi\_type field is added with the aim of distinguishing with ease (just a INT flag) whether the row of interest is relative to a green (value 1) or a yellow cab (value 2).

### Feeding the working table "rides\_lite"

```
INSERT OVERWRITE TABLE rides_lite
SELECT
YEAR(lpep_pickup_datetime) as year,
RatecodeID,
PULocationID,
DOLocationID,
passenger_count,
trip_distance,
total_amount,
CASE WHEN (REVERSE(SPLIT(REVERSE(INPUT__FILE__NAME), '[/]')[0])) LIKE
'green%' THEN 1 WHEN (REVERSE(SPLIT(REVERSE(INPUT__FILE__NAME),
'[/]')[0])) LIKE 'yellow%' THEN 2 END
FROM rides;
```

- Noticing that the objective queries require only the year of a ride, it's been chosen to insert only this inside the rides\_lite table using the function YEAR which extracts the year from a string date. This expedient lightens the computational and memory requirements, also because the year is here treated as an integer instead of a string.
- To value the field taxi\_type of the table rides\_lite, Hive checks if the file name starts with the keyword green or yellow:
  - The keyword INPUT\_\_FILE\_\_NAME is a virtual column [5] which contains the full path of the source file related to that row.
  - The algorithm consists of reversing the full path and getting the first word delimited by a slash (resulting to be the file name with reversed letters disposition), extracting the first element and reversing it again to get the file name of the CSV file; then, depending by the type of the taxi, Hive assigns 1 or 2 to the taxi type field using the CASE statement and the LIKE operator.

INPUTFILENAME	hdfs://localhost:9000/user/hive/warehouse/assignment_taxi.db/rides/green_tripdata_2016-07.csv
REVERSE	vsc.70- 6102_atadpirt_neerg/sedir/bd.ixat_tnemngissa/esuoheraw/evih/resu/000 9:tsohlacol//:sfdh
SPLIT on /	[0 => vsc.70-6102_atadpirt_neerg] [1 => sedir] [2 => bd.ixat_tnemngissa] [3 => esuoheraw] [4 => evih] [5 => resu] [6 => 0009:tsohlacol] [7 => ] [8 => :sfdh]
SPLIT_ELEMENT[0]	vsc.70-6102_atadpirt_neerg
REVERSE	green_tripdata_2016-07.csv

### **DQL** commands

To store the result in a file it is possible to put before the query, the following command:

```
INSERT OVERWRITE LOCAL DIRECTORY 'home/result hive'
```

### Query 4

```
SELECT year, SUM(passenger_count) FROM rides_lite GROUP BY year HAVING
year = '2018';
```

• This query uses the SUM function to sum the *passenger\_count* and group the results by year.

### Query 5

```
SELECT RatecodeID, SUM(total_amount) FROM rides_lite WHERE year = '2018'
GROUP BY RatecodeID;
```

This query uses the SUM function to sum the *total\_amount*, for rides in 2018 and group the results by RatecodeID.

### Query 6

```
SELECT PULocationID, COLLECT_SET(DOLocationID) as group_doloc, COLLECT_LIST(avg_trip) as group_trip FROM (SELECT PULocationID, DOLocationID, CAST(FORMAT_NUMBER(AVG(trip_distance), 2) AS FLOAT) AS avg_trip FROM rides_lite WHERE taxi_type = 1 AND year = '2018' GROUP BY PULocationID, DOLocationID ORDER BY avg_trip) table_temp GROUP BY PULocationID;
```

This query is made up of two queries:

- Internal query: selecting all the yellow taxi (taxi\_type = 1), it gets the values of the columns *PULocationID* and *DOLocationID* and groups the average of *trip\_distance* by *PULocationID*. The average results are expressed in miles, approximated to the second decimal digit, for ease of reading purposes. The results are sort by average trip.
- External query: groups by *PULocationID* and display the results of *DOLocationID* and avg trip in a "list-like" mode.

# **Dealing with different datasets**

To make the measurement comparison three datasets have been created:

- A table containing data for 2018, 2017 e the second half of 2016, inside the table *rides\_lite*.
- A table containing data for 2018 and 2017 only, called *rides\_lite\_2018\_2017*.
- A table containing data from 2018 only, called *rides\_lite\_2018*.

### Construction and population of the table for the years 2018 and 2017

```
CREATE TABLE rides lite 2018 2017 (
year INT,
RatecodeID INT,
PULocationID INT,
DOLocationID INT,
passenger count INT,
trip distance FLOAT,
total amount INT,
taxi type INT
);
INSERT OVERWRITE TABLE rides lite 2018 2017
YEAR (lpep pickup datetime) as year,
RatecodeID,
PULocationID,
DOLocationID,
passenger count,
trip distance,
total amount,
CASE WHEN (REVERSE(SPLIT(REVERSE(INPUT FILE NAME), '[/]')[0])) LIKE
'green%' THEN 1 WHEN (REVERSE(SPLIT(REVERSE(INPUT FILE NAME),
'[/]')[0])) LIKE 'yellow%' THEN 2 END
FROM rides WHERE lpep pickup datetime LIKE '2018%' OR lpep pickup datetime
LIKE '2017%';
```

### Construction and population of the table for the year 2018

```
CREATE TABLE rides_lite_2018 (
year INT,
RatecodeID INT,
PULocationID INT,
DOLocationID INT,
passenger_count INT,
trip_distance FLOAT,
total_amount INT,
taxi_type INT
);
INSERT OVERWRITE TABLE rides lite 2018
```

```
SELECT
YEAR(lpep_pickup_datetime) as year,
RatecodeID,
PULocationID,
DOLocationID,
passenger_count,
trip_distance,
total_amount,
CASE WHEN (REVERSE(SPLIT(REVERSE(INPUT__FILE__NAME), '[/]')[0])) LIKE
'green%' THEN 1 WHEN (REVERSE(SPLIT(REVERSE(INPUT__FILE__NAME),
'[/]')[0])) LIKE 'yellow%' THEN 2 END
FROM rides WHERE lpep pickup datetime LIKE '2018%';
```

### Query 4 for the table rides\_lite\_2018\_2017

SELECT year, SUM(passenger\_count) FROM rides\_lite\_2018\_2017 GROUP BY year
HAVING year = '2018';

### Query 5 for the table rides\_lite\_2018\_2017

SELECT RatecodeID, SUM(total\_amount)FROM rides\_lite\_2018\_2017 WHERE year =
'2018' GROUP BY RatecodeID;

### Query 6 for the table rides\_lite\_2018\_2017

SELECT PULocationID, COLLECT\_SET(DOLocationID) as group\_doloc, COLLECT\_LIST(avg\_trip) as group\_trip FROM (SELECT PULocationID, DOLocationID, CAST(FORMAT\_NUMBER(AVG(trip\_distance), 2) AS FLOAT) AS avg\_trip FROM rides\_lite\_2018\_2017 WHERE taxi\_type = 1 AND year = '2018' GROUP BY PULocationID, DOLocationID ORDER BY avg\_trip) table\_temp GROUP BY PULocationID;

### Query 4 for the table rides\_lite\_2018

SELECT year, SUM(passenger count) FROM rides lite 2018 GROUP BY year;

### Query 5 for the table rides\_lite\_2018

SELECT RatecodeID, SUM(total\_amount)FROM rides\_lite\_2018 GROUP BY
RatecodeID;

### Query 6 for the table rides\_lite\_2018

SELECT PULocationID, COLLECT\_SET(DOLocationID) as group\_doloc, COLLECT\_LIST(avg\_trip) as group\_trip FROM (SELECT PULocationID, DOLocationID, CAST(FORMAT NUMBER(AVG(trip distance), 2) AS FLOAT) AS

avg\_trip FROM rides\_lite\_2018 WHERE taxi\_type = 1 GROUP BY PULocationID,
DOLocationID ORDER BY avg trip) table temp GROUP BY PULocationID;

# **Dealing with different operational modes**

Another parameter to take in account when it comes to performance analysis is the use of the hive operational modes: local (exectution on the local machine as a simple node) and pseudo-distributed (simulation of a node in a network)

### Activation of the local mode

```
SET mapred.job.tracker = local;
SET hive.exec.mode.local.auto = true;
```

### Activation of the pseudo-distributed node

```
SET mapred.job.tracker = pseudo;
SET hive.exec.mode.local.auto = false;
```

# Results

# Query 4

Year	SUM(passenger_count)
2018	16789109123

# Query 5

RatecodeID	SUM(total_amount)
3	135431705
35	393328
67	16894
99	4002
131	601
163	498
195	1433
547	0
611	44
18	38855698
50	87956
82	5783
114	2455
146	350
178	493
210	486
274	811
6098	10000
10	48361500
42	151954
74	10863
106	1746
138	466
20	16510505
52	78108

84 5190	84
116 1195	116
148 593	148
212 240	212
2 228543473	2
34 373812	34
66 28521	66
98 3858	98
130 1291	130
162 300	162
226 700	226
258 10005	258
290 729	290
6 49935919	6
38 257534	38
70 18206	70
102 1438	102
134 456	134
166 190	166
198 400	198
230 52	230
422 1059	422
25 1728785	25
57 42296	57
89 5796	89
121 931	121
153 1047	153
4 87835275	4
36 442198	36
68 18582	68
100 3300	100
132 1498	132
164	164
196 983	196

484	2417
932	14
1764	4414
12	24786499
44	135621
76	9504
108	685
140	763
204	1598
300	7
24	1983564
56	42497
88	3436
120	3028
152	277
216	302
248	412
600	5
5	70676335
37	344676
69	20014
101	10318
133	311
229	450
325	50
1061	8
2053	5134
5381	10001
22	5743867
54	59485
86	4615
118	3398
150	275
182	25

471	214
1322	246
584971	31
27367	63
3141	95
1363	127
782	159
11	831
41743771	7
225538	39
16419	71
3766	103
635	135
10001	7655
490008478	1
418931	33
27326	65
3820	97
2854	129
12	801
3 23	833
1648840	28
35305	60
4264	92
4166	124
66	156
3 10	188
1340	252
0	380
10000	6204
30579674	16
90516	48
4983	80
2 1801	112

144	510
176	2658
528	1325
4016	10000
19	23170653
51	75338
83	3399
115	1923
147	1051
211	1227
1459	3652
1811	4530
26	1874646
58	39006
90	4830
122	1150
154	462
218	338
250	656
602	18
1402	3509
1818	4551
17	49407742
49	90402
81	6218
113	2657
145	700
753	1887
1297	3247
1841	4606
0 2	06982960
32	539941
64	27540
96	5126

1353	128
1720835	27
36994	59
4318	91
1727	123
300	155
1177	251
26	603
4311	1723
15348514	14
101087	46
7740	78
2605	110
2405	142
581	206
25	270
234	302
300	910
2897899	23
56069	55
3805	87
2192	119
1822	151
0	183
5581	2231
52519319	9
172225	41
10075	73
1912	105
1021	137
372	169
0	329
16976114	13
119482	45

77	8688
109	3146
141	374
301	25
11 37	619022
43	137584
75	6425
107	2154
139	1077
267	950
189483	4
29	.044709
61	33477
93	3641
125	1484
157	225
189	170
15 18	3439856
47	98777
79	8051
111	2853
207	502
239	639
943	14
21 11	125701
53	61243
85	5157
117	3677
149	394
181	875
213	858
245	466
341	1002
8 47	251101

188208	40
12932	72
2613	104
298	136
211	168
4824	1928
6023	2408
710537	30
30268	62
3368	94
1366	126
C	222
31	606
9	830

# Query 6 (first row only)

PULocati onID	COLLECT_SET(DOLocationID) AS group_doloc	COLLECT_LIST(avg_trip) AS group_trip
1	[1,265,23,249,66,68,14,164,22,243,14 5,166,257,64,132]	[1.06,8.93,12.7,12.86,15.95,17.1,18.71,20.1,22.0,23.9,25.21,2 6.64,32.9,34.9,37.1,161.19]

# **Performance comparison**

The program instances are run on the following virtual machine:

The operative system is Ubuntu 18.4.4 hosted by VirtualBox vers. 6.0.14 which are given:

2 cores; 4096 MB RAM; 60GB disk space;

Equipped with Hadoop vers. 2.8.5, Java Open JDK vers. 1.8.0

The host machine with:

Intel©  $Core^{TM}$  i7 6500U and SSD.

### Time for completion

years: 2018

number of entries in table rides\_lite\_2018: 111609853

Task	Execution time in local mode (s)	Execution time in psuedo-distributed (s)
Query 4	118.137	131.078
Query 5	129.997	127.181
Query 6	134.667	147.617

years: 2017, 2018

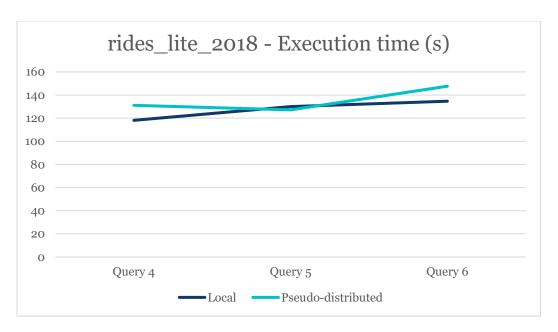
number of entries in table rides\_lite\_2018\_2017: 236847300

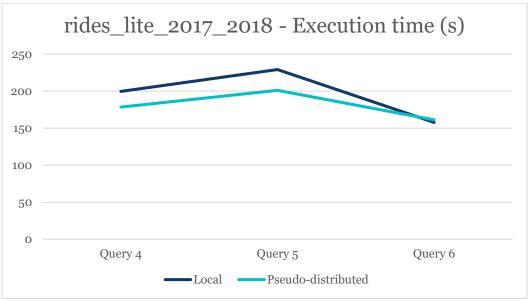
Task	Execution time in local mode (s)	Execution time in psuedo-distributed (s)
Query 4	199.514	178.525
Query 5	229.105	201.061
Query 6	157.725	161.184

years: 2016, 2017, 2018

number of entries in table *rides\_lite*: 305975113

Task	Execution time in local mode (s)	Execution time in psuedo-distributed (s)
Query 4	354.845	211.382
Query 5	297.944	243.932
Query 6	409.915	525.028







The pseudo-distributed mode performs better or worse than the local mode, depending on the type of query and the amount of data: the biggest differences are noticeable in case of the whole dataset table (made up of over 300M rows) where the pseudo-distributed mode takes -40.42 % time in comparison to local mode in terms of completion time for the query 4, while it uses +28.08 % time more when the query 6 is run.

### **Limitations**

The results concerning the execution times are biased by the state of the host machine which was loaded with different tasks in background, causing a not constant access time and bandwidth availability on the SSD on the host operating system, so these must be taken as an overall idea of the average execution time for the machine used.

### References

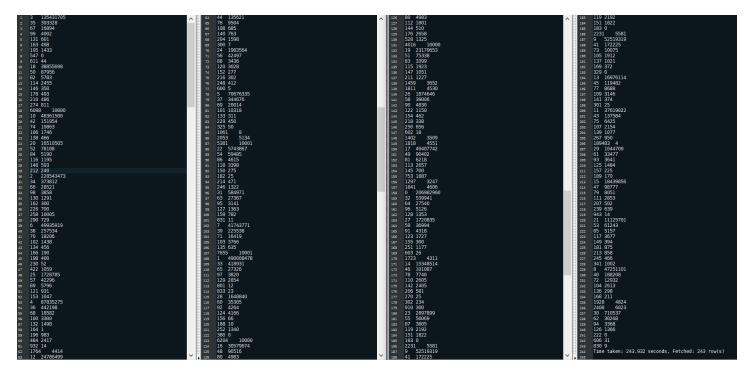
- [1] https://en.wikipedia.org/wiki/New York City Taxi and Limousine Commission
- [2] https://www.quora.com/What-is-the-difference-between-Green-Cabs-and-Yellow-Cabs
- [3] https://www1.nyc.gov/assets/tlc/downloads/pdf/data\_dictionary\_trip\_records\_yellow.pdf
- [4] https://www1.nyc.gov/assets/tlc/downloads/pdf/data dictionary trip records green.pdf
- [5] https://cwiki.apache.org/confluence/display/Hive/LanguageManual+VirtualColumns

# **Appendix**

### Screenshot of resulting file for Query 4

```
1 2018 16789109123
2 Time taken: 131.078 seconds, Fetched: 1 row(s)
```

### Screenshot of resulting file for Query 5



### Screenshot of resulting file for the Query 6 (first row only)

1 [1,265,23,249,66,68,14,164,22,243,145,166,257,64,132] [1.06,8.93,12.7,12.86,15.95,17.1,18.71,20.1,22.0,23.9,25.21,26.64,32.9,34.9,37.1,161.19]

### Row composition for Query 6 file

Red: PULocationID

Light blue: list of DOLocationID

Green: sorted list of trip distance in ascending order relative to DOLocationID