

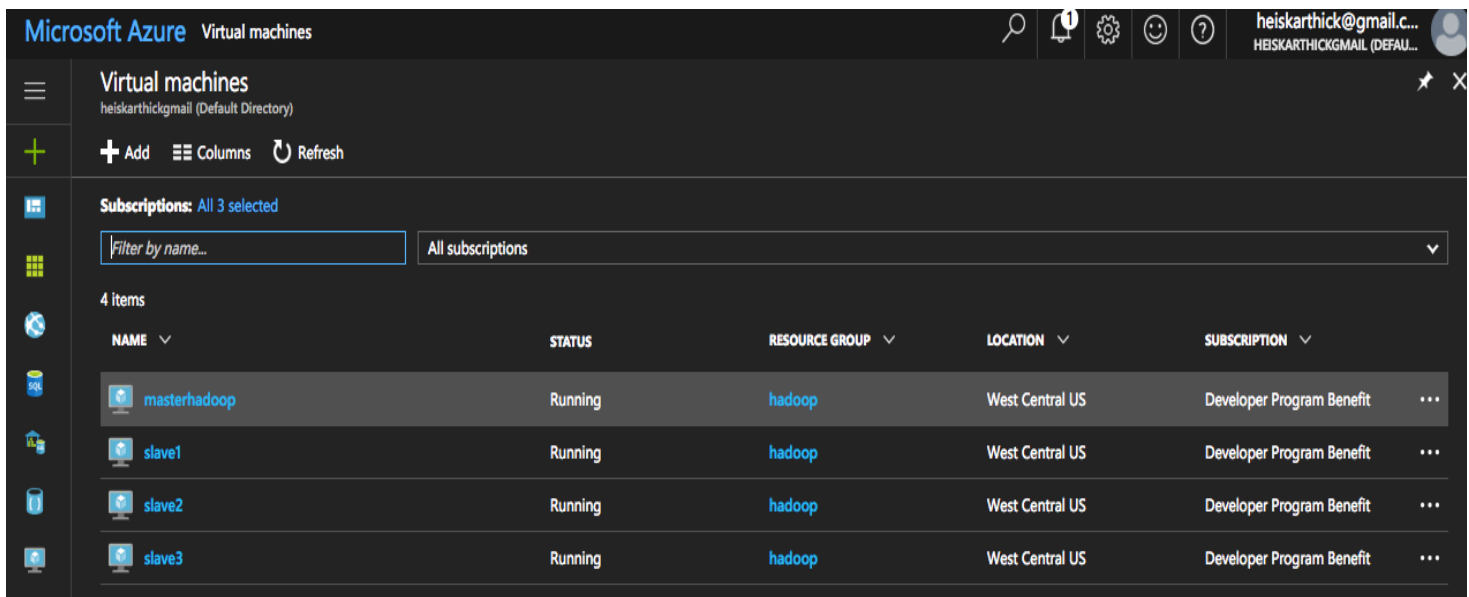
Introduction:

The aim of this project is to set up a HDFS cluster in Microsoft azure and to run MapReduce program to process on Yellow Cab data, from New York City Taxi And Limousine Commission dataset which include fields capturing pick-up and drop-off dates/times, pick-up and drop-off locations, trip distances, itemized fares, rate types, payment types, and driver-reported passenger counts.

System behaviour:

1. This system consists of a master and 3 clones of the master called as slaves.
2. This cluster has one name node and 3 data nodes with a replication factor of 3.
3. Start the Hadoop deamons.
4. Download the Yellow Cab data from New York City Taxi in master which is then pushed to datanode in all 3 slaves.
5. MapReduce program is written in python language to run a map reduce function on the downloaded dataset.

Cluster formation



The screenshot displays the Microsoft Azure portal interface for Virtual machines. The page title is "Virtual machines" under the user "heiskarthick@gmail (Default Directory)". The interface includes a sidebar with navigation icons and a main content area. In the main area, there are buttons for "Add", "Columns", and "Refresh". Below these, it shows "Subscriptions: All 3 selected" and a search bar "Filter by name...". A dropdown menu shows "All subscriptions". Below this, it states "4 items" and displays a table of virtual machines.

NAME	STATUS	RESOURCE GROUP	LOCATION	SUBSCRIPTION
masterhadoop	Running	hadoop	West Central US	Developer Program Benefit
slave1	Running	hadoop	West Central US	Developer Program Benefit
slave2	Running	hadoop	West Central US	Developer Program Benefit
slave3	Running	hadoop	West Central US	Developer Program Benefit

Description of MapReduce program:

There are two separate python files called as map and reduce in master node.

Pre-processing the dataset in Map.py

The first 5 months data of Yellow Cab NYC in the year 2010 is taken for this project.

1. Firstly, Removing the headers and blank space is done at the beginning as it is required to remove these in the .csv data files.
2. Remove the white spaces that are there in the file.
3. separate each word as tokens that are separated by comma (,).
4. Take the vendor name separately at line 0, passenger count at line 3, trip distance at line 4, fare amount at line 12, payment at line 11 and date at line 10.
5. Finally, Create dictionaries to store the mapped data of each vendor according to the trip distance, fare amount etc.

The dictionaries are:

- No. of trips per vendor
The trip distance will be appended to this dictionary at every iteration
- Fare amount vendor
Fare amount of each vendor will be appended to this dictionary
- Payment types
We append the total number of times each payment method was used per vendor
- Months in year
Month will be the key and the trip distance per month will be kept in this dictionary
- Total no. of passengers per vendor
We keep the passenger count per vendor here

Reduce.py

Reducer takes all the key, value pairs from mapper, aggregates and generates the desired output.

Data Analysis:

17/05/02 07:36:50 WARN util.NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-java classes where applicabl

DDS 993839.433333

VTs 7763327.31159

CMT 6982015.00025

DDS 3524072.47583

VTs 28151799.4600

CMT 26130411.8033

credit 16813730

no charge 221254

cash 53873315

dispute 42517

2009-05 2.74477415833

2009-04 2.68741308432

2009-03 2.70218479269

2009-02 2.60981118353

2009-01 2.55777528823

DDS 5662497.0

VTs 72135578.0

CMT 42102840.0

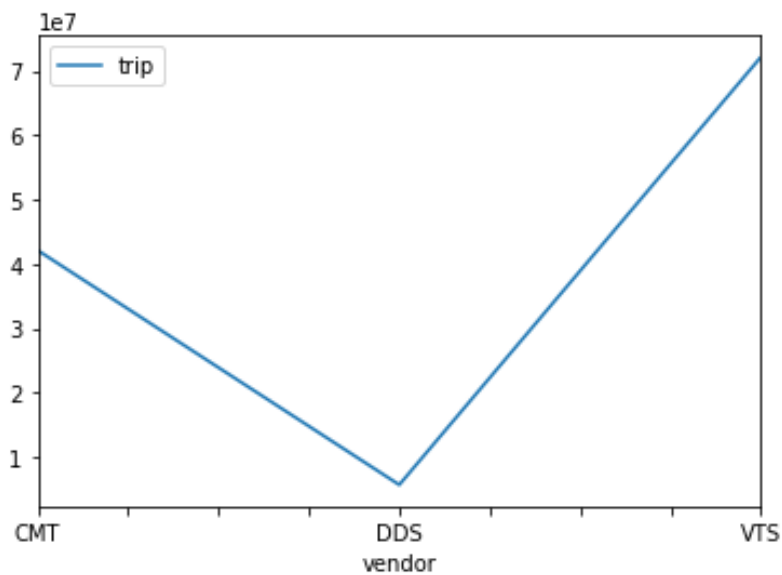
The mostly used payment type

Cash	53873315
Credit	16813730
No charge	221254
Dispute	42517

The table shows that cash was majorly used as payment type in the year 2010.

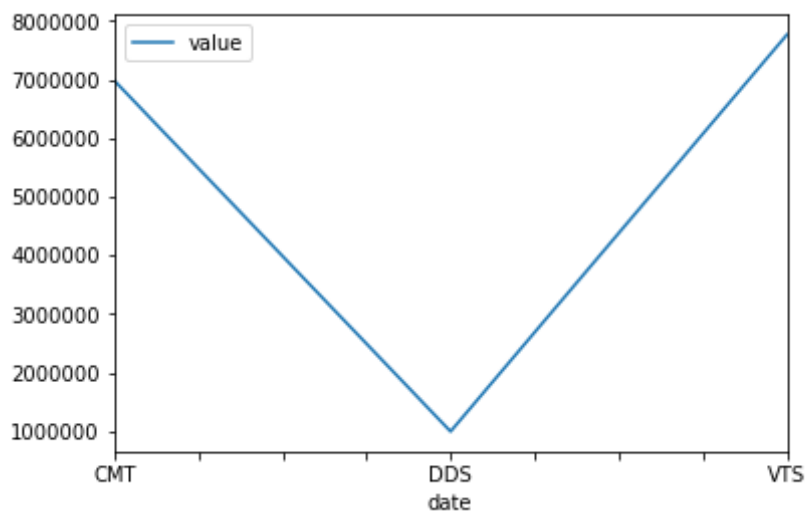
No. of trips took by each vendor in 2010

DDS	5662497.0
VTs	72135578
CMT	42102840



Total trip distance by each vendor by each month

DDS	993839.433333
VTs	7763327.31159
CMT	6982015.08025

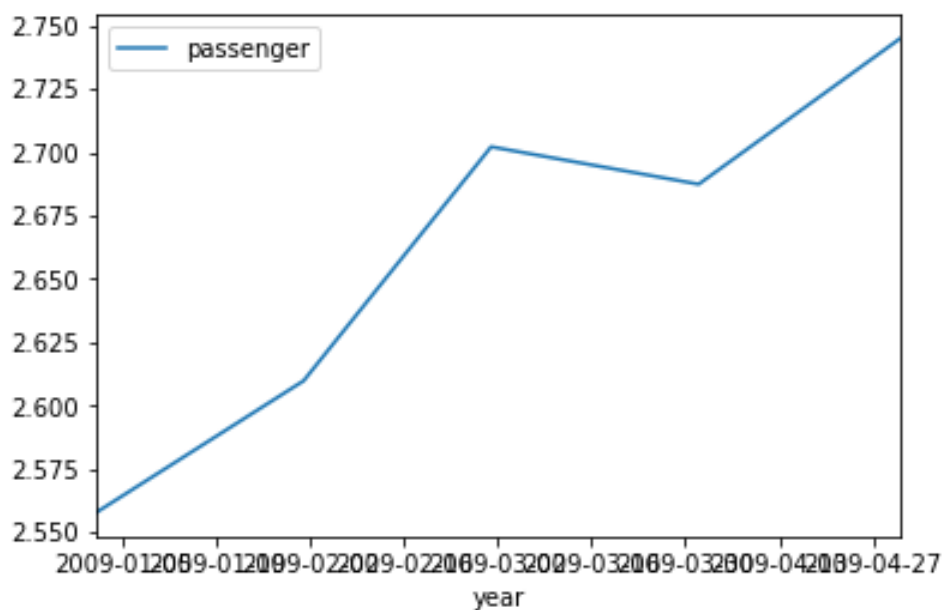


Total fare by each vendor Jan to May in 2010

DDS	3524072.47583
VTs	28151799.4608
CMT	26130411.8033

Average no. of passengers per month per day

2009-05	2.74477415833
2009-04	2.68741308432
2009-03	2.70218479269
2009-02	2.60981118353
2009-01	2.55777528823



Challenges faced and its solution:

1. Creating a cluster:

- The master node's port number had to be removed from the hdfs-site.xml as this led to an error when I tried to SSH the master node and name node did not start because it was in the master node.

- The disks were found that they could be downloaded to the local HDD if the **“ActiveSAS” state** was enabled. But this led to an error when the clusters were created and had to be **unchecked**.

2. Memory limitation:

As the vms have limited memory, It was challenging to download all data and pushed into cluster. Therefore, limited data is downloaded and pushed into cluster for data analysis.

Access to VMs:

Masterhadoop	ssh hadoop@ 13.78.179.87
Slave1	ssh hadoop@ 52.161.8.221
Slave2	ssh hadoop@ 52.161.29.205
Slave3	ssh Hadoop@ 52.161.8.140

Command to run:

```
hadoop jar /usr/local/Cellar/hadoop/2.8.0/libexec/share/hadoop/tools/lib/hadoop-streaming-2.8.0.jar -mapper map.py -reducer reducer.py -input /2010 -output /2010output -file map.py -file reducer.py
```

Conclusion:

Having worked in this project, I gained valued in-sights in creating vms and setting up a cluster to run MapReduce program to process datasets from NYC.