Project outline and used technologies

The project dedicates to the processing stream and non stream data on the technologies that enable by using the same architecture processing and analyzing Big Data. For the detailed description of dataset and project problem statement please refer to the pdf file

ProblemStatement_Process_and_collect_GPS_trajectory_dataset.pdf

This document is dedicated to guiding through the code base of the project and point on the places on the code base where was implemented the solution to the problem which was imposed in ProblemStatement_Process_and_collect_GPS_trajectory_dataset.pdf for this purpose I integrate the marks like <<impl of solution to 3.0.1 ps>> which basically mean that in the nearest print screen you can find precise place where the solution of the subproblem was implemented.

At the end of this document, you can find the hyperlink to the video demos of data flow and jar output that prove in some degree that indeed the codebase can produce the needed result without bugs or exceptions. On the next page, you will find the architecture of the system with specification outline that I use the project solution flow. All solution was made on the customer made ubuntu OS (VM) with all installations steps which I do not describe in the document. If you will need some elaboration u can always contact me for further comments or consultations.

The components and technologies which was used:

OS Windows 10 - as a host OS for GPSDataStreamSimulator

OS Ubuntu 16.04 - used as a host OS for hadoop cluster

Apache Hadoop 2.7.5 - used as a host cluster for HBase

Apache Maven 3.3.9 - used as building tool for GPSDataStreamSimulator

SBT 1.01 - build tool for the shc-df.jar and rt-spark.jar

Apache Zookeeper 3.4.10 - used as a main synchronization between HBase , messaging broker Kafka, hadoop in general and etc

Apache Kafka 2.11-1.0.0 - used as a broker messaging system

Apache Spark 2.1.1-hadoop2.7- used for data analysis and writing to the HBase

Apache HBase 1.1.12 - NoSQL database for fast store and retrieval of data

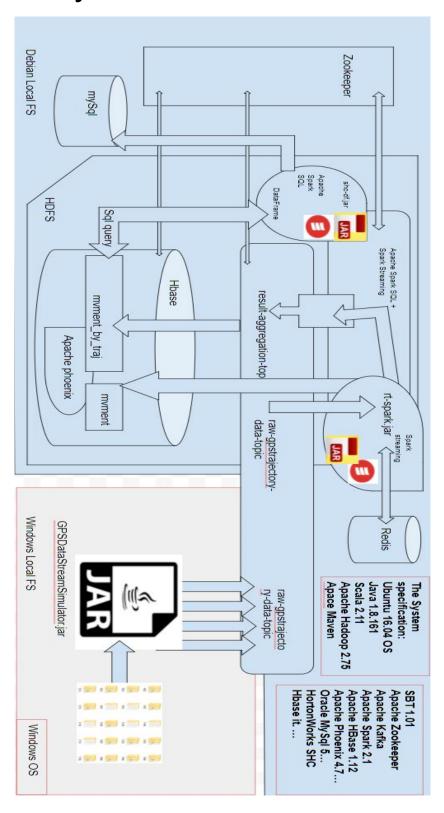
Apache Phoenix 4.8.1 - for the monitoring and fast data analysis

Redis 4.0.8 - as a buffer component for calculating "dist" column and "tdiff"

Oracle MySql 5.7.19 - for post analysis storate of the analysis result

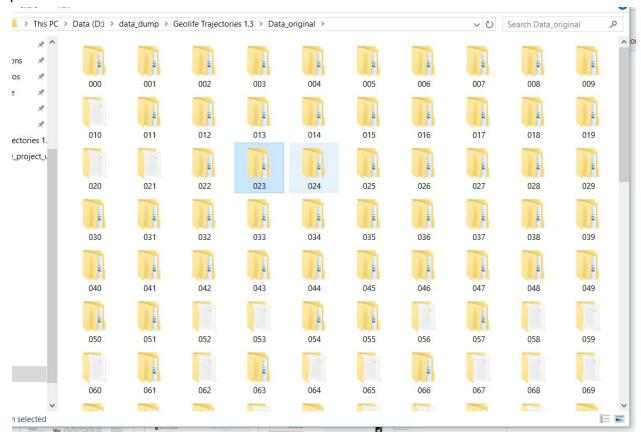
HortonWorks SHC 1.1.2-2.2-s_2.11 - for connection between apache spark and HBase to enable retrieving data as a DataFrame and enable running apache spark data analysis on the top of the YARN

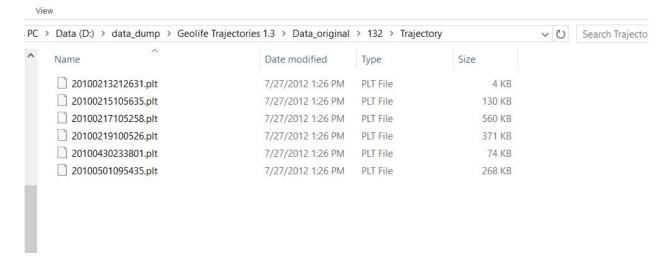
2. System Architecture description



3. Data Ingestion and initial Validation

The original data contain 181 folders each correspond to the user in which several trajectories presented





So the first challenge that we should overcome is to somehow merge all these files into 1 - 8 corresponding how many cores we wanna dedicate to streaming data. Also in the big files the

chronological order should be maintained as in original one in order to make possible to deliver data enrichment in our system.

```
D:\data_dump\Geolife Trajectories 1.3\Data_original\132\Trajectory\20100219100526.plt - Notepad++
File Edit Search View Encoding Language Settings Tools Macro Run Plugins Window ?
 3 🔒 🗎 🖺 🤚 🧸 🥱 🚵 🔏 🛍 🖍 🖿 🖿 🖊 🗷 🖒 🕩 🕞 🗩 🖒 🗀 🕦 🖊
🔚 20100219100526.plt 🔀
     29.880386,121.603702,0,0,40228.425625,2010-02-19,10:12:54
    29.880371,121.603528,0,0,40228.4256481482,2010-02-19,10:12:56
    29.880359,121.603348,0,0,40228.4256712963,2010-02-19,10:12:58
    29.880358,121.603166,0,0,40228.4256944444,2010-02-19,10:13:00
    29.880357,121.60298,0,0,40228.4257175926,2010-02-19,10:13:02
     29.880358,121.602791,0,0,40228.4257407407,2010-02-19,10:13:04
     29.880356,121.602581,0,0,40228.4257638889,2010-02-19,10:13:06
     29.880349,121.602374,0,0,40228.425787037,2010-02-19,10:13:08
     29.880338,121.602148,0,0,40228.4258101852,2010-02-19,10:13:10
    29.880332,121.601927,0,0,40228.4258333333,2010-02-19,10:13:12
    29.880332,121.601705,0,0,40228.4258564815,2010-02-19,10:13:14
    29.880332,121.601489,0,0,40228.4258796296,2010-02-19,10:13:16
    29.880333,121.601278,0,0,40228.4259027778,2010-02-19,10:13:18
    29.880323,121.601068,0,0,40228.4259259259,2010-02-19,10:13:20
    29.880324,121.60085,0,0,40228.4259490741,2010-02-19,10:13:22
     29.880329,121.600638,0,0,40228.4259722222,2010-02-19,10:13:24
     29.88033,121.600404,0,0,40228.4259953704,2010-02-19,10:13:26
     29.880333,121.600161,0,0,40228.4260185185,2010-02-19,10:13:28
     29.880334,121.599938,0,0,40228.4260416667,2010-02-19,10:13:30
     29.880327,121.599751,0,0,40228.4260648148,2010-02-19,10:13:32
     29.880321,121.599604,0,0,40228.426087963,2010-02-19,10:13:34
    29.88033,121.599483,0,0,40228.4261111111,2010-02-19,10:13:36
    29.880337,121.599369,0,0,40228.4261342593,2010-02-19,10:13:38
     29.88034,121.599231,0,0,40228.4261574074,2010-02-19,10:13:40
     29.88034,121.599065,0,0,40228.4261805556,2010-02-19,10:13:42
     29.880343,121.598872,0,0,40228.4262037037,2010-02-19,10:13:44
     29.88035,121.598678,0,0,40228.4262268519,2010-02-19,10:13:46
     29.88035,121.598475,0,0,40228.42625,2010-02-19,10:13:48
     29.880347,121.598279,0,0,40228.4262731481,2010-02-19,10:13:50
     29.88035,121.598074,0,0,40228.4262962963,2010-02-19,10:13:52
    29.880343,121.597878,0,0,40228.4263194444,2010-02-19,10:13:54
    29.880346,121.597685,0,0,40228.4263425926,2010-02-19,10:13:56
     29.880349,121.597475,0,0,40228.4263657407,2010-02-19,10:13:58
     29.880347,121.597241,0,0,40228.4263888889,2010-02-19,10:14:00
     29.880351.121.596999.0.0.40228.426412037.2010-02-19.10:14:02
Normal text file
                                                                    length: 379,604 lines: 6
```

In order to solve the problem described above, we create the module in our project that will be called GPSDataSimulation which will contain three classes: KafkaMassageProducer, DataSourcesRunner, FileMerger, SeparateDataSourcesSimulator and the first that we will be FileMerger.

FileMerger

class will take several functions:

- 1. read data from all 181 folders
- 2. Validate the data, filter invalid pieces of data and write it in log file

<< Impl of sol 3.0.4 ps >>

3. Sort all data to maintain chronological order

4. Merge all 181 with more than 780+ files into (by default) four separate big files, which will be ready to put in into broker messaging pipeline for further processing and storage

KafkaMessageProducer

class will be used for:

1. Contain all necessary configuration for the message broker system that will be used to consume data for the further processing.

The Apache Kafka broker was chosen because of it guaranty preservation of chronological order of each message delivery which valuable for our use case.

SeparateDataSourceSimulator

class will be used for :

1. To read each line of the file

- 2. Produce all necessary logs to control and monitor data flow
- 3. Write data (line by line) in the Kafka broker

```
LOG.log(Level.INFO, msg: "inside loop before send ");
this.kafkaMessageProducer.send(this.topic, line);
LOG.log(Level.INFO, msg: "inside loop ");
if (i.incrementAndGet() % 5 == 0) {
LOG.log(Level.INFO, msg: "sending lines in progress. " +
```

To avoid CPU overhead we simulate random latency between consuming each piece of data to Kafka broker

```
if (i.incrementAndGet() % 5 == 0) {
    LOG.log(Level.INFO, msg: "sending lines in progress. " + "Check point: " + i);
}

try {
    //noinspection AccessStaticViaInstance
    Thread.currentThread().sleep(this.random.nextInt( bound: 222));
} catch (InterruptedException ex) {
    Logger.getLogger(SeparateDataSourceSimulator.class.getName()).log(Level.SEVERE, msg)
}
```

DataSourcesRunner

Finally, the DataSourcesRunner class will be used for :

- 1. Point to the folder from which the system should take the data and put it in apache Kafka broker
- 2. Point to the name of the topic that will produce GPS raw data flow after initial validation and preparation
- Start up several (by default number of threads equals a number of files in the data directory) several threads that will simulate separate data sources that will write to broker messaging system.

```
<<impl of solution to 3.0.1 ps>>
```

<<iinpl of solution to 3.0.2 ps>>

VM configuration

Choosing component version

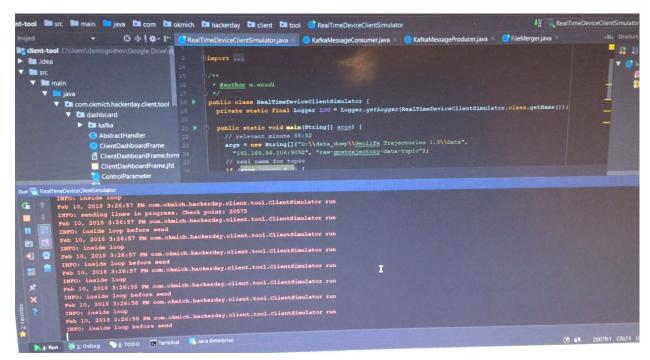
In order to meet all necessary architectural requirements and avoid incompatibility of the version of the components let's refer to the industry leaders product and choose the versions of the

component for example like in Hortonworks HDP. So in hortonworks.com, we can find all necessary specification of HDP 2.6 and HDP 2.5 and we'll keep in mind all these versions while building our own VM distribution.

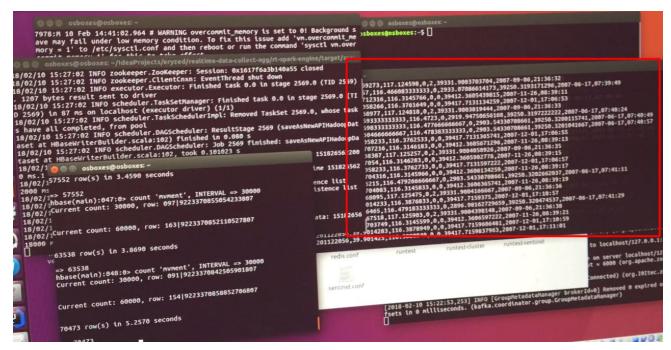


Push data stream to Apache Kafka

In order to check that the data successfully could put and retrieve from our messaging broker, we can create console consumer in VM ubuntu OS and take a look does data appears in terminal window or not



Where the data that is pushing to message broker



Here the same data as u can see but on the other side of the broker (push out to the console window)

Now we can move to another step is that pushing data stream to HBase.

4. Data Enrichment

Push data stream to HDFS Database

- 1. We need to create tables in HBase
 - a. For this purpose, we will use HBase shell
 - b. We need to create 'mvment' table with column family 'main' for the raw valid data (to satisfy problem statement requirements << *** type the requirement point >>
 - c. We create 'mvment_by_traj' table with column family 'main' in which enrichment data will be collected and against which we will make our analysis (from point 5 the case problem statement)
- 2. Now we need to push out data from broker message to apache spark streaming
- 3. In Apache Spark (rt-spark.jar) we will make:
 - a. Further validation and enrichment todo >> add name of class
 - b. Push data stream (one micro batch at the single iteration) to HBase todo >> add name of class

```
| Trisspark-engine | Instrumental |
```

```
<< Impl of sol 3.0.3 ps >>
```

- << Impl of sol 3.0.5 ps >>
- << Impl of sol 4.1.1 ps >>
- << Impl of sol 4.1.2 ps >>
- << Impl of sol 4.1.3 ps >>
- << Impl of sol 5.1.2 ps >>

Because trajectory model contain "userld" field and all necessary fields for making advanced analytics of data database by could completely avoid usage of such expensive operation as a SQL query << JOIN>>

```
ark-engine src main scala @ Main.scala
                             rt-spark-engine ~/IdeaProjects/hbase_a
                                                                       * @param rdd
project [rt-spark-engine-build] soun
                                                                   def processRDD(rdd: RDD[String]): Unit = {
                                                                      val readingRDD: RDD[Reading] = rdd.map(createReading(_))
   ▼ ■ scala
                                                                      //cache rdd for preadingRDD.cache
                                                                      val trajectoryMoveMentRDD: RDD[Trajectory] = readingRDD.mapPartitions { partitionOfRecords => {
                                                                     val trajectoryNovementRUD: RUD[[rajectory] = readingRUD.mapPartitions { partition
val cache = new CacheService
val rdds = partitionOfRecords.map((reading: Reading) => {
val plat = cache.getAndSet("lt" + reading.trajectoryId, reading.lat.toString)
val plon = cache.getAndSet("ln" + reading.trajectoryId, reading.tos.toString)
val pts = cache.getAndSet("ts" + reading.trajectoryId, reading.ts.toString)
       ▼ I model
       ▼ 🖿 redis
         @ Main
                                                                          Trajectory(if (plat == null) reading.lat else plat.toFloat,
   if (plon == null) reading.lon else plon.toFloat,
                                                                                reading.lat, reading.lon,
if (pts == null) reading.ts else pts.tolong,
reading.ts, reading.trajectoryId, reading.userId)
 build.sbt
shc_df [shc-examples] ~/IdeaProjects/b
                                                                      //save raw data to hbase saveReadingToHBase(readingRDD)
                                                                       saveTrajectoryToHBase(trajectoryMoveMentRDD)
```

```
<< Impl of sol 4.1.4 ps >>
<< Impl of sol 4.2.1 ps >>
<< Impl of sol 4.2.2 ps >>
<< Impl of sol 5.1.1 ps >>
<< Impl of sol 5.1.5 ps >>
```

Data flow monitoring and initial analysis

If we use just four threads to simulated 4 difference GPS devices with random latency (0-200 ms) the transfer of 2 Gb dataset that initially was given to us to transfer to hdfs and substantially analyze will take a lot of time.

In order to make sure that during this process the data indeed comes to HBase, we need to adopt at least a simple monitor tools that alert as in case the system failure. For this purpose, we will be used :

- 1. Hbase shell
- 2. Apache Phoenix

So this two tools will show us that indeed the amount of data in HBase is increasing during the transfer and validation is apache spark stream going well

```
02/10 15:34:34 INFO SCHEDULE DAGSCHEDU
                                                                                                             20071211041037
                                                                                                                                           604.1353791717804
t at HBaseWriterBuilder.scala:102) finished in
                                                                                                             20071211055512
02/10 15:34:34 INFO scheduler.DAGScheduler: Job
et at HBaseWriterBuilder.scala:102, took 0.0983
                                                                                                                                           1697.474412312787
                                                                                                             20071212094726
                                                                                     O Recent
                                                                                                                                           12526.205985952374
                                                                                                             28871215015200
                                                                                                                                           6135.219517073533
                                                                                                                          124912
02/1 😞 🖨 📵 osboxes@osboxes: ~
                                                                                                                                           5272.076471602152
                                                                                                                           101100
ns.1
ns.1

/02/1

00 ms=> 82706

/02/1hbase(main):050:0> count 'mvment', INTERVAL => 30000

/02/1current count: 30000, row: 076|9223370860097689807

3/02/1current count: 60000, row: 140|9223370858781591807

3/02/183055 row(s) in 4.9840 seconds
                                                                                                                                           2791.16988073774
                                                                                                                           195300
                                                                                                                                           18808.617766164854
                                                                                                                          022509
                                                                                                                                           6254.297919670056
                                                                                                                           103108
                                                                                                                                           1458.1696139364233
                                                                                                                          055038
                                                                                                                                           340.33457191560547
                                                                                                                           041235
                                                                                                                                           7823.519804230325
                                                                                                                           124344
8/02/1
8/02/1

8/02/1=> 83055

70000 rhbase(main):051:0> count 'mvment_by_traj', INTERVAL => 30000

70000 rhbase(main):051:0> count 'mvment_by_traj', INTERVAL => 30000

7 Current count: 30000, row: 20070608232055|9223370848255752807
                                                                                                                                                   DISTANCE
                                                                                                                           id
                                                                                                                                           1177.7260361866113
         **Current count: 60000, row: 20070901022340|9223370848255752807
V<sup>E</sup>77825 row(s) in 8.4060 seconds
                                                                                                                           092307
                                                                                                                                           11014.919196652832
                                                                                                                           060101
                                                                                                                                           10137.11072217434
                                                                                                                           023337
                                                                                                                                           21222.064793492682
                                                                                                                           042937
                                                                                                                                          6831.979881877159
            hbase(main):052:0> count 'mvment_by_traj', INTERVAL => 30000
Current count: 30000, row: 20070608232055|9223370855519261807
Current count: 60000, row: 20070901022340|9223370848255648807
                                                                                                                           063113
                                                                                                                           elected (0.419 seconds)
                                                                                                                           oenix:localhost>
              78182 row(s) in 7.2660 seconds
                                                                                                                              user by trajectories
              hbase(main):053:0> count 'mvment_by_traj', INTERVAL => 30000
Current count: 30000, row: 20070608232055|9223370855519261807
Current count: 60000, row: 20070828171302|9223370848544303807
                                                                                                                           elect "userId", count(distinct "

    user by collection period
elect "userId", min("ts") mints,
```

<< Impl of sol 5.1.3 ps >>

5. Data analysis

Now will query the data by using apache spark sql that connects to HBase through hortonworks connector for building further more advanced analysis of the data

To check that indeed all connectivity between apache spark SQL and HBase going well we can run the same query in apache phoenix. In case of receiving the different result from the same query, it will alert us about some system malfunction. So good to have apache phoenix shell or squirrel GUI ready for work

Because we use HortonWorks hbase - spark connector it enables to run the apache spark job on the top of the YARN which is important as a part of the assignment 5.1.4

```
<< Impl of sol 5.1.4 ps >>
```

- << Impl of sol 5.0 ps >>
- << Impl of sol 5.0.1 ps >>
- << Impl of sol 5.0.2 ps >>
- << Impl of sol 5.0.3 ps >>
- << Impl of sol 5.0.4 ps >>
- << Impl of sol 5.0.5 ps >>

```
### Light Summer Summer
```

6. Post Analysis

Transfer and save analytical result to RDBMS database

<< Impl of sol 6.0.1 ps >>

Appendix:

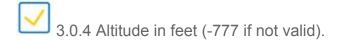
Problems outline

3. Data Ingestion and Initial Validation



3.0.2 All the timestamp fields in data coming from data sources are of the format YYYY-MM-DD HH:MM:SS.

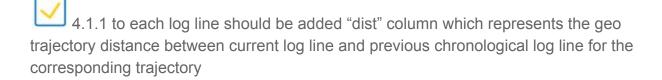
3.0.3 Finally, all timestamps must have the format of a long integer to be interpreted as UNIX timestamps when they reach database in HDFS.



3.0.5 Create an identifier for each data line.

4. Data Enrichment

4.1 Rules for data enrichment



4.1.2 to each log line should be calculated "tdiffs" column which represents the time difference between current log line and previous chronological log line for the corresponding trajectory

4.1.3 if the previous timestamp of corresponding trajectory not older than current timestamp of the trajectory set the time difference is 0

4.1.4 each calculation should be performed in real-time or near real-time mode

4.2 Post Enrichment



4.2.2 Move all valid records in HBase

5. Data analysis



- 5.0.2 Determine top 10 users amount of trajectories.
- 5.0.3 Determine top 10 trajectories by duration.
- 5.0.4 Determine top 10 endured trajectory and which user they belong

5.1. Challenges and optimization

5.1.1 Raw data and processed data comes to NoSQL database as a stream. All calculations and validations performed as a data streams.



5.1.3 Adopt appropriate monitoring to maintained to track the behavior and overcome failures in the pipeline.

5.1.4 For data analytics in spark use technologies that could be work with Hadoop resource manager like YARN.



6. Post Analysis

6.0.1. Design functionally to enable result of analytics move to the RDBMS for data storage and quick retrieval.

Appendix 2:

Video demos:

https://streamable.com/7axck - running_sql_pnoenix_spark
https://streamable.com/jqdt3 - demo_tranfer_from_win_to_ubuntu_via_kafka
https://streamable.com/ks5ru - 50k_lines_of_data_monitoring_pnoenix_hbase_shell
https://streamable.com/wiaqa - 100k_lines_of_data_monitoring_pnoenix_hbase_shell
https://streamable.com/p3xoi - 150k_lines_of_data_monitoring_pnoenix_hbase_shell
https://streamable.com/d9ebv - 200k_lines_of_data_monitoring_pnoenix_hbase_shell
https://streamable.com/6n6o4 - 300k_lines_of_data_monitoring_pnoenix_hbase_shell