INDUSTRIAL TRAINING REPORT - II

Done by

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At

National Informatics Centre- Open Technology Group. Software Technology Parks of India, Taramani Chennai- 90.

Submitted to



Department of Computer Science Engineering Faculty of Engineering and Technology SRM University (Under section 3 of UGC Act, 1956) SRM Nagar, Kattankulathur – 603203 Kanchipuram District.

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BONAFIDE CERTIFICATE

Certified that this Industrial Training Report – II is a work of ROHITH.V (Reg.No.RA1411003010245) who carried out the work at NATIONAL INFORMATICS CENTRE- OPEN TECHNOLOGY GROUP, Software Technology Parks of India, Taramani, Chennai, Tamil Nadu.

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Year-Coordinator

(Mrs. PUSHPALATHA)

Certificate



भारत सरकार / Government of India Ministry of Electronics & Information Technology (MeitY) इलेक्ट्रॉनिकी और सूचना प्रौद्योगिकी मंत्रालय

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Ref.No. NIC - OTG /HND/ 41 2017.

Date: 28/06/2017

R.RAMESH

Technical Director

TO WHOMSOEVER IT MAY CONCERN

This is to certify that Mr.ROHITH.V (RA1411003010245) a student from B.TECH. (Computer Science Engineering) has successfully completed his Industrial Training in National Informatics Centre (NIC) from 05th to 28th June 2017 on the title "Sentiment Analysis using Twitter and Apache Spark".

The student had showed keen interest and curiosity in learning new things and Wish him all success in his future endeavours.

28/6/2017 R.RAMESH

Acknowledgement

It is always a pleasure to remind the fine people in the Engineering program for their sincere guidance I received to uphold my practical as well as theoretical skills in engineering.

Firstly I would like to thank Dr. C.Muthamizhchelvan (Director of E&T, SRM University) for meticulously planning academic curriculum in such a way that students are not only academically sound but also industry ready by including such industrial training patterns.

I would also like to thanks Mr. C.Rajesh Babu (Class Incharge) for the positive attitude he showed for my work, always allowing me to question him and giving prompt replies for my uncertainties in all the fields including educational, social and managerial work.

I express my immense pleasure and deep sense of gratitude to Mr. R.Ramesh (Technical Director) and to my guide Ms.Agalya (Developer) for spending her valuable time with me and also helped me in completion of task.

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Company Profile:



The **National Informatics Centre (NIC)** is the premier science & technology organisation of India's Union Government in informatics services and information-and-communication- technology (ICT) applications. The NIC is a part of the Indian Ministry of Communications and Information Technology's Department of Electronics & Information Technology.

It has played a pivotal role in steering e-governance applications in the governmental departments at national, state and district levels, enabling the improvement in, and a wider transparency of, government services. Almost all Indian-government websites are developed / managed by NIC.

Field of work

NIC offers telecommunications-networking services including K_u band (TDMA, FDMA, SCPC & satellite broadband) VSATs, wireless metropolitan-area networks (MANs) and local-area networks (LANs) with gateways for Internet- and Intranet-resource sharing.

It is the network infrastructure and e-governance support to India's central government and state governments, union-territory administrations, administrative divisions and other government bodies. The NIC assists in implementing information-technology projects, in collaboration with central and state governments, in the areas of: Communication & Information Technology.

Organisation

NIC is a part of the Indian Ministry of Communications & Information Technology's Department of Electronics & Information Technology and is headquartered in New Delhi. It has offices in all 29 state capitals and 7 union-territory headquarters and almost all districts. At New Delhi Headquarters, Mean Head a large number of Application Divisions exist which provide total Informatics Support to the Ministries and Departments of the Central Government. To cater to the ICT needs at the grassroots level, the NIC has also opened offices in almost all district collectorates. NIC Extends Technical Coordination and IT support to District Administration.

Services

The NIC offers a host of services including:

- Computer-aided design (CAD)
- Geographical-Information system (GIS)
- Domain-Name Registration for gov.in and nic.in domain
- Informatics
 - Biomedical informatics
 - Patent informatics
 - Rural informatics
 - o Agriculture informatics, including hydrography
- Internet Data Centre (IDC)
- Mathematical Modelling and Simulation
- Computer networking
- office-procedure automation (OPA)
- Training
- Cyber Security
- Videoconferencing
- Website hosting & website development
- Internet Services
- Wi-Fi

Open Technology Group

NIC has established Open Technology Group (OTG) to spearhead the technology exploration and provision support services for adoption of OSS in various e-Governance Projects and applications under NIC and NeGP program of DeitY. NIC-OTG is mandated to facilitate strategic control of Open Technology within NIC and spearhead the knowledge centric activities in e-Governance Projects all over India.

The Open Technology Group (OTG), National Informatics Centre (NIC) has implemented the Open Technology Centre Project (OTC) (A Grant-in-Aid Project) of Department of Electronics & (DeitY), Government of India (GoI) during the period (April 2007- March 2015).

Key Technology Services supported by OTG

- 1. CMS/Portal using Drupal
- 2. Open Web Platform
- 3. Database Replication using Symmetric DS
- 4. Database Migration to PostgreSQL
- 5. Single Sign on Solution using CAS
- 6. Verification Services based on 2D Barcode
- 7. Platform independent Digital Signature Certificate
- 8. Recommendation of Open Source Stack after due evaluation
- 9. Bundled OSS Stack for Development, Staging & Deployment
- 10. Capacity Building & Hand holding on Recommended Open Source Stack
- 11. Provisioning Support for recommended stack-List of empanelled Vendors-Quick Reference
- 12.Performance Tuning of Open Source Application Servers

Also the Group participates in the Open Source related activities of NeST project of DeitY implemented by NICSI with STQC as technology partner.

Platforms used:

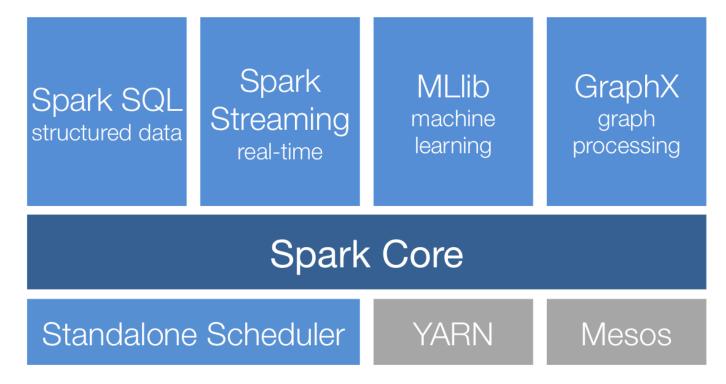
1.	Apache Spark
2.	Apache Open NLP Classifier
3.	Eclipse IDE
4.	Scala
5.	Java
6.	Maven Repository
7.	Twitter API

APACHE SPARK:



Apache Spark is an open-source cluster-computing framework. Originally developed at the University of California, Berkeley's AMPLab, the Spark codebase was later donated to the Apache Software Foundation, which has maintained it since. Spark provides an interface for programming entire clusters with implicit data parallelism and fault-tolerance.

Overview



Apache Spark provides programmers with an application programming interface centered on a data structure called the resilient distributed dataset (RDD), a read-only multiset of data items distributed over a cluster of machines, that is maintained in a fault-tolerant way. It was developed in response to limitations in the MapReduce cluster computing paradigm, which forces a particular linear dataflow structure on distributed programs: MapReduce programs read input data from disk, map a function across the data, reduce the results of the map, and store reduction results on disk. Spark's RDDs function as a working set for distributed programs that offers a

(deliberately) restricted form of distributed shared memory.

The availability of RDDs facilitates the implementation of both iterative algorithms, that visit their dataset multiple times in a loop, and interactive/exploratory data analysis, i.e., the repeated database-style querying of data. The latency of such applications (compared to a MapReduce implementation, as was common in Apache Hadoop stacks) may be reduced by several orders of magnitude. Among the class of iterative algorithms are the training algorithms for machine learning systems, which formed the initial impetus for developing Apache Spark.

Apache Spark requires a cluster manager and a distributed storage system. For cluster management, Spark supports standalone (native Spark cluster), Hadoop YARN, or Apache Mesos. For distributed storage, Spark can interface with a wide variety, including Hadoop Distributed File System (HDFS), MapR File System (MapR-FS), Cassandra, OpenStack Swift, Amazon S3, Kudu, or a custom solution can be implemented. Spark also supports a pseudo-distributed local mode, usually used only for development or testing purposes, where distributed storage is not required and the local file system can be used instead; in such a scenario, Spark is run on a single machine with one executor per CPU core.

Spark Core

Spark Core is the foundation of the overall project. It provides distributed task dispatching, scheduling, and basic I/O functionalities, exposed through an application programming interface (for Java, Python, Scala, and R) centered on the RDD abstraction). This interface mirrors a functional/higher-order model of programming: a "driver" program invokes parallel operations such as map, filter or reduce on an RDD by passing a function to Spark, which then schedules the function's execution in parallel on the cluster. These operations, and additional ones such as joins, take RDDs as input and produce new RDDs. RDDs are immutable and their operations are lazy; fault-tolerance is achieved by keeping track of the "lineage" of each RDD (the sequence of operations that produced it) so that it can be reconstructed in the case of data loss. RDDs can contain any type of Python, Java, or Scala objects.

Spark SQL

It is a component on top of Spark Core that introduced a data abstraction called DataFrames, which provides support for structured and semi-structured data. Spark SQL provides a domain-specific language (DSL) to manipulate DataFrames in Scala, Java, or Python. It also provides SQL language support, with command-line interfaces

and ODBC/JDBC server. Although DataFrames lack the compile-time type-checking afforded by RDDs, as of Spark 2.0, the strongly typed DataSet is fully supported by Spark SQL as well.

Spark Streaming

Spark Streaming leverages Spark Core's fast scheduling capability to perform streaming analytics. It ingests data in mini-batches and performs RDD transformations on those mini-batches of data. This design enables the same set of application code written for batch analytics to be used in streaming analytics, thus facilitating easy implementation of lambda architecture. However, this convenience comes with the penalty of latency equal to the mini-batch duration. Other streaming data engines that process event by event rather than in minibatches include Storm and the streaming component of Flink. Spark Streaming has support built-in to consume from Kafka, Flume, Twitter, ZeroMQ, Kinesis, and TCP/IP sockets.

MLlib Machine Learning Library

Spark MLlib is a distributed machine learning framework on top of Spark Core that, due in large part to the distributed memory-based Spark architecture, is as much as nine times as fast as the disk-based implementation used by Apache Mahout (according to benchmarks done by the MLlib developers against the Alternating Least Squares (ALS) implementations, and before Mahout itself gained a Spark interface), and scales better than Vowpal Wabbit. Many common machine learning and statistical algorithms have been implemented and are shipped with MLlib which simplifies large scale machine learning pipelines, including:

- summary statistics, correlations, stratified sampling, hypothesis testing, random data generation
- classification and regression: support vector machines, logistic regression, linear regression, decision trees, naive Bayes classification
- collaborative filtering techniques including alternating least squares (ALS)
- cluster analysis methods including k-means, and Latent Dirichlet Allocation (LDA)
- dimensionality reduction techniques such as singular value decomposition (SVD), and principal component analysis (PCA)
- feature extraction and transformation functions optimization algorithms such as stochastic gradient descent, limited-memory BFGS (L-BFGS)

GraphX

GraphX is a distributed graph processing framework on top of Apache Spark. Because it is based on RDDs, which are immutable, graphs are immutable and thus GraphX is unsuitable for graphs that need to be updated, let alone in a transactional manner like a graph database. GraphX provides two separate APIs for implementation of massively parallel algorithms (such as PageRank): a Pregel abstraction, and a more general MapReduce style API. Unlike its predecessor Bagel, which was formally deprecated in Spark 1.6, GraphX has full support for property graphs (graphs where properties can be attached to edges and vertices). GraphX can be viewed as being the Spark in-memory version of Apache Giraph, which utilized Hadoop disk based MapReduce. Like Apache Spark, GraphX initially started as a research project at UC Berkeley's AMPLab and Databricks, and was later donated to the Apache Software Foundation and the Spark project.

APACHE OPEN NLP:



The Apache OpenNLP library is a machine learning based toolkit for the processing of natural language text. It supports the most common NLP tasks, such as tokenization, sentence segmentation, part-of-speech tagging, named entity extraction, chunking, parsing, and co-reference resolution. These tasks are usually required to build more advanced text processing services. OpenNLP also included maximum entropy and perceptron based machine learning.

The goal of the OpenNLP project will be to create a mature toolkit for the abovementioned tasks. An additional goal is to provide a large number of prebuilt models for a variety of languages, as well as the annotated text resources that those models are derived from.

General Library Structure

The Apache OpenNLP library contains several components, enabling one to build a full natural language processing pipeline. These components include: sentence detector, tokenizer, name finder, document categorizer, part-of-speech tagger, chunker, parser, co reference resolution. Components contain parts which enable one to execute the respective natural language processing task, to train a model and often also to evaluate a model. Each of these facilities is accessible via its application program interface (API). In addition, a command line interface (CLI) is provided for convenience of experiments and training.

ECLIPSE IDE:



Eclipse is an integrated development environment (IDE) used in computer programming, and is the most widely used Java IDE. It contains a base workspace and an extensible plug-in system for customizing the environment. Eclipse is written mostly in Java and its primary use is for developing Java applications, but it may also be used to develop applications in other programming languages via plug-ins, including Ada, ABAP, C, C++, COBOL, D, Fortran, Haskell, JavaScript, Julia, Lasso, Lua, NATURAL, Perl, PHP, Prolog, Python, R, Ruby (including Ruby on Rails framework), Rust, Scala, Clojure, Groovy, Scheme, and Erlang. It can also be used to develop documents with LaTeX (via a TeXlipse plug-in) and packages for the software Mathematica. Development environments include the Eclipse Java development tools (JDT) for Java and Scala, Eclipse CDT for C/C++, and Eclipse PDT for PHP, among others.

The initial codebase originated from IBM VisualAge. The Eclipse software development kit (SDK), which includes the Java development tools, is meant for Java developers. Users can extend its abilities by installing plug-ins written for the Eclipse Platform, such as development toolkits for other programming languages, and can write and contribute their own plug-in modules. Since Equinox, plug-ins can be plugged-stopped dynamically and are termed (OSGI) bundles.

Eclipse software development kit (SDK) is free and open source software, released under the terms of the Eclipse Public License, although it is incompatible with the GNU General Public License. It was one of the first IDEs to run under GNU Classpath and it runs without problems under IcedTea.

SCALA:



Scala is a general-purpose programming language providing support for functional programming and a strong static type system. Designed to be concise, many of Scala's design decisions aimed to address criticisms of Java.

Scala source code is intended to be compiled to Java bytecode, so that the resulting executable code runs on a Java virtual machine. Scala provides language interoperability with Java, so that libraries written in both languages may be referenced directly in Scala or Java code. Like Java, Scala is object-oriented, and uses a curly-brace syntax reminiscent of the C programming language. Unlike Java, Scala has many features of functional programming languages like Scheme, Standard ML and Haskell, including currying, type inference, immutability, lazy evaluation, and pattern matching. It also has an advanced type system supporting algebraic data types, covariance and contravariance, higherorder types (but not higher-rank types), and anonymous types. Other features of Scala not present in Java include operator overloading, optional parameters, named parameters, raw strings, and no checked exceptions.

The name Scala is a portmanteau of scalable and language, signifying that it is designed to grow with the demands of its users.

JAVA:



Java is a general-purpose computer programming language that is concurrent, class-based, object-oriented, and specifically designed to have as few implementation dependencies as possible. It is intended to let application developers "write once, run anywhere" (WORA),meaning that compiled Java code can run on all platforms that support Java without the need for recompilation.[16] Java applications are typically compiled to bytecode that can run on any Java virtual machine (JVM) regardless of computer architecture. As of 2016, Java is one of the most popular programming languages in use, particularly for client-server web applications, with a reported 9 million developers. Java was originally developed by James Gosling at Sun Microsystems (which has since been acquired by Oracle Corporation) and released in 1995 as a core component of Sun Microsystems' Java platform. The language derives much of its syntax from C and C++, but it has fewer low-level facilities than either of them.

TWITTER:



Twitter is an online news and social networking service where users post and interact with messages, "tweets", restricted to 140 characters. Registered users can post tweets, but those who are unregistered can only read them. Users access Twitter through its website interface, SMS or a mobile device app. Twitter Inc. is based in San Francisco, California, United States, and has more than 25 offices around the world.

Developers

Twitter is recognized for having one of the most open and powerful developer APIs of any major technology company. Developer interest in Twitter began immediately following its launch, prompting the company to release the first version of its public API in September 2006. The API quickly became iconic as a reference implementation for public REST APIs and is widely cited in programming tutorials.

From 2006 until 2010, Twitter's developer platform experienced strong growth and a highly favorable reputation. Developers built upon the public API to create the first Twitter mobile phone clients as well as the first URL shortener. Between 2010 and 2012, however, Twitter made a number of decisions that were received unfavorably by the developer community. In 2010, Twitter mandated that all developers adopt OAuth authentication with just 9 weeks of notice. Later that year, Twitter launched its own URL shortener, in direct competition with some of its most well-known 3rd-party developers. And in 2012, Twitter introduced strict usage limits for its API, "completely crippling" some developers. While these moves successfully increased the stability and security of the service, they were broadly perceived as hostile to developers, causing them to lose trust in the platform.

In an effort to reset its relationship with developers, Twitter acquired Crashlytics on January 28, 2013 for over USD \$100 million, its largest acquisition to date. Founded by Jeff Seibert and Wayne Chang, Crashlytics had rapidly gained popularity as a tool to

help mobile developers identify and fix bugs in their apps. Twitter committed to continue supporting and expanding the service.

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In early 2016, Twitter announced that Fabric was installed on more than 2 billion active devices and used by more than 225,000 developers. Fabric is recognized as the #1 most popular crash reporting and also the #1 mobile analytics solution among the top 200 iOS apps, beating out Google Analytics, Flurry, and MixPanel.

NECESSITY:

Spark Streaming: Real time twitter sentiment analysis

Spark Streaming is an extension of the core Spark API that enables scalable, high-throughput, fault-tolerant stream processing of live data streams. Spark Streaming can be used to stream live data and processing can happen in real time. Spark Streaming's evergrowing user base consists of household names like Uber, Netflix and Pinterest.

When it comes to Real Time Data Analytics, Spark Streaming provides a single platform to ingest data for fast and live processing in Apache Spark. Through this blog, I will introduce you to this new exciting domain of Spark Streaming and we will go through a complete use case, *Twitter Sentiment Analysis* using Spark Streaming.

Spark Streaming API can consume from sources like Kafka, Flume, and Twitter source to name a few. It can then apply transformations on the data to get the desired result which can be pushed further downstream.

What is Streaming?

Data Streaming is a technique for transferring data so that it can be processed as a steady and continuous stream. Streaming technologies are becoming increasingly important with the growth of the Internet.

Why Spark Streaming?

We can use Spark Streaming to stream real-time data from various sources like Twitter, Stock Market and Geographical Systems and perform powerful analytics to help businesses.

Spark Streaming Overview

Spark Streaming is used for processing real-time streaming data. It is a useful addition to the core Spark API. Spark Streaming enables high-throughput and fault-tolerant stream processing of live data streams.

The fundamental stream unit is DStream which is basically a series of RDDs to process the real-time data.

Spark Streaming Features

- 1. Scaling: Spark Streaming can easily scale to hundreds of nodes.
- 2. Speed: It achieves low latency.
- 3. Fault Tolerance: Spark has the ability to efficiently recover from failures.
- 4. *Integration:* Spark integrates with batch and real-time processing.
- 5. *Business Analysis:* Spark Streaming is used to track the behavior of customers which can be used in business analysis.

Spark Streaming Workflow

Spark Streaming workflow has four high-level stages. The first is to stream data from various sources. These sources can be streaming data sources like Akka, Kafka, Flume, AWS or Parquet for real-time streaming. The second type of sources includes HBase, MySQL, PostgreSQL, Elastic Search, Mongo DB and Cassandra for static/batch streaming. Once this happens, Spark can be used to perform Machine Learning on the data through its MLlib API. Further, Spark SQL is used to perform further operations on this data. Finally, the streaming output can be stored into various data storage systems like HBase, Cassandra, MemSQL, Kafka, Elastic Search, HDFS and local file system.

Spark Streaming Fundamentals

- 1. Streaming Context
- 2. DStream
- 3. Caching
- 4. Accumulators, Broadcast Variables and Checkpoints

Streaming Context

Streaming Context consumes a stream of data in Spark. It registers an *Input DStream* to produce a *Receiver* object. It is the main entry point for Spark functionality. Spark provides a number of default implementations of sources like Twitter, Akka Actor and ZeroMQ that are accessible from the context.

A StreamingContext object can be created from a SparkContext object. A SparkContext represents the connection to a Spark cluster and can be used to create RDDs, accumulators and broadcast variables on that cluster.

Example:

```
import org.apache.spark._
import org.apache.spark.streaming._
var ssc = new StreamingContext(sc,Seconds(1))
```

DStream

Discretized Stream (DStream) is the basic abstraction provided by Spark Streaming. It is a continuous stream of data. It is received from a data source or a processed data stream generated by transforming the input stream.

Internally, a DStream is represented by a continuous series of RDDs and each RDD contains data from a certain interval.

Input DStreams:

Input DStreams are DStreams representing the stream of input data received from streaming sources.

Every input DStream is associated with a Receiver object which receives the data from a source and stores it in Spark's memory for processing.

Transformations on DStreams:

Any operation applied on a DStream translates to operations on the underlying RDDs. Transformations allow the data from the input DStream to be modified similar to RDDs. DStreams support many of the transformations available on normal Spark RDDs.

The following are some of the popular transformations on DStreams:

map(func)	map(func) returns a new DStream by passing each element of the source DStream through a function func.
flatMap(func)	flatMap(<i>func</i>) is similar to map(<i>func</i>) but each input item can be mapped to 0 or more output items and returns a new DStream by passing each source element through a function <i>func</i> .
filter(func)	filter(<i>func</i>) returns a new DStream by selecting only the records of the source DStream on which <i>func</i> returns true.

reduce(func)	reduce(<i>func</i>) returns a new DStream of single-element RDDs by aggregating the elements in each RDD of the source DStream using a function <i>func</i> .
groupBy(func)	groupBy(<i>func</i>) returns the new RDD which basically is made up with a key and corresponding list of items of that group.

Output DStreams:

Output operations allow DStream's data to be pushed out to external systems like databases or file systems. Output operations trigger the actual execution of all the DStream transformations.

Caching

DStreams allow developers to cache/ persist the stream's data in memory. This is useful if the data in the DStream will be computed multiple times. This can be done using the *persist()* method on a DStream.

For input streams that receive data over the network (such as Kafka, Flume, Sockets, etc.), the default persistence level is set to replicate the data to two nodes for fault-tolerance.

Problem Statement

To design a Twitter Sentiment Analysis System where we populate real-time sentiments for crisis management, service adjusting and target marketing.

Applications of Sentiment Analysis:

- Predict the success of a movie
- Predict political campaign success
- Decide whether to invest in a certain company
- Targeted advertising
- Review products and services

Architecture:



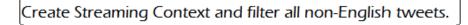
WORKFLOW:

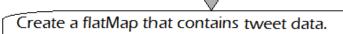
Spark Popular HashTags

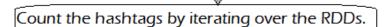


Import all necessary dependencies from Maven Repository and include them in the project.

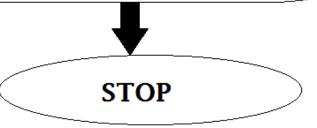




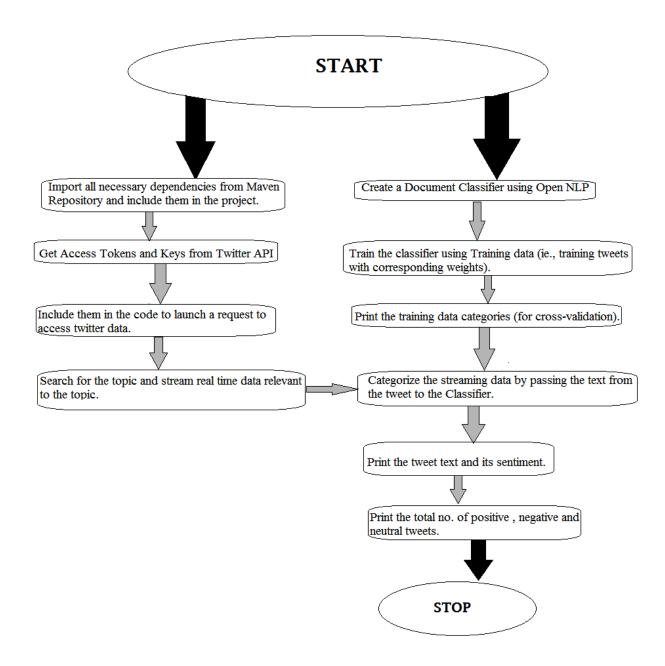




Print the hashtags along with its count.

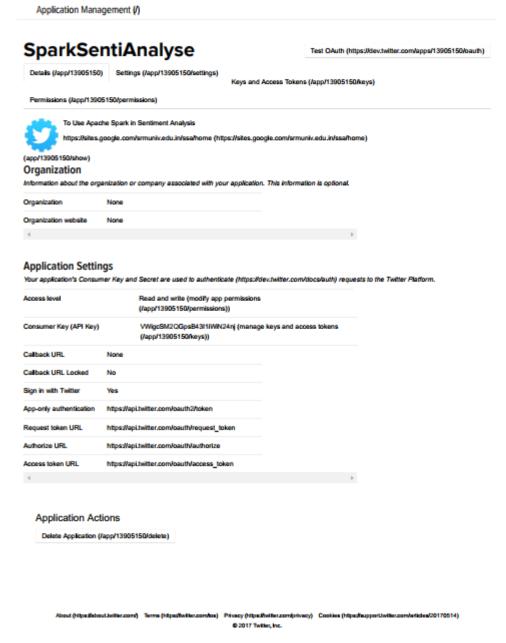


Sentiment Analysis with count:



Twitter API:

i. Log on to https://www.apps.twitter.com/ with Twitter Credentials.



Set the app name and the website URL. ii.

Application Manage	ement (/)	
SparkSen	tiAnalyse	Test OAuth (https://dev.twitter.com/apps/13905150/oauth)
Details (/app/13905150)	Settings (/app/13905150/settings)	Keys and Access Tokens (/app/13905150/keys)
Permissions (/app/139051	50/permissions)	
Application Detail	ils	
Name *		
SparkSentiAnalyse		
Your application name. This	is used to attribute the source of a tweet and	in user-facing authorization acreena. 32 characters max.
Description *		
To Use Apache Spark in	n Sentiment Analysis	
Your application description,	which will be shown in user-facing authoriza	tion screens. Selecen 10 and 200 characters max.
Website *		
	/srmuniv.edu.in/ssa/home	
		download, make use of, or find out more information about your application. This fully-qualified URL is used in the
	created by your application and will be show	
(Fyou don't have a URL yet,	just put a placeholder here but remember to	change it later.)
Caliback URL		
		polities, with example applications should explicitly specify their courts_caliback URL on the request token step,
regardless of the value given	here. To restrict your application from using	calibacks, leave this field blank.
Privacy Policy URL		
The URL for your application	or service's privacy policy. The URL will be a	shared with users authorizing this application.
Terms of Service URL		
The URL for your application	or service's terms of service. The URL will b	e shared with users authorizing this application.
E Souble College's control	in a fit is an annumented to an abit of the	and building to account account account to the collins of
	* 1	eack locking to ensure apps cannot overwrite the caliback urf)

iii. Take note of the Consumer Key, Consumer Token, Access Key and Access token in the Keys

Application Setti Keep the **Consumer Se	ngs cret" a secret. This key should never be human-readable in your application.
Consumer Key (API Key	
Consumer Secret (API S	secret)
Access Level	Read and write (modify app permissions (/app/13905150/permissions))
Owner	Morra_Gambit
Owner ID	
	umer Key and Secret (/app/13905150/recreate_keys) issions (/app/13905150/permissions)
Your Access Tok	en
This access token can b	e used to make API requests on your own account's behalf. Do not share your access token secret with anyone.
Access Token	
Access Token Secret	
Access Level	Read and write
Owner	Morra_Gambit
Owner ID	

iv. Set app permissions, if necessary.

SparkSentiAnalyse	Test OAuth (https://dev.twitter.com/apps/13905150/oau
Details (/app/13905150) Settings (/app/13905150/settings)	
Keys an	d Access Tokens (/app/13905150/keys)
Permissions (/app/13905150/permissions)	
Access	
What type of access does your application need?	
Read more about our Application Permission Model (https://dev.twitter.com	m/docs/application-permission-model).
Read only	
Read and Write	
 Read, Write and Access direct messages 	
Note:	
Changes to the application permission model will only reflect in access t	tokens obtained after the permission model change is saved. You will
need to re-negotiate existing access tokens to alter the permission level	I associated with each of your application's users.
Additional Permissions	
These additional permissions require that you provide URLs to your applithese fields in your Application Settings (/app/13905150/settings).	cation or service's privacy policy and terms of service. You can configure
triese fields in your Application Settings (rapp/13903130/settings).	

CODE:

Maven Dependencies:

```
WithCount.java 🗵 📙 OpenNLPCategorizer.java 🗵 📘 SparkTwitterDataProcessor.java 🗵 🔜 sparkit.java 🗵 📑 pom.xml 🗵
             kproject xmlns="http://maven.apache.org/POM/4.0.0" xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance xsi:schemaLocation="http://maven.apache.org/POM/4.0.0 http://maven.apache.org/xsd/maven-4.0.0.xsd">
http://maven.apache.org/POM/4.0.0 http://maven.apache.org/xsd/maven-4.0.0.xsd">
http://maven.apache.org/POM/4.0.0 http://maven.apache.org/xsd/maven-4.0.0.xsd">
http://maven.apache.org/POM/4.0.0 http://maven.apache.org/xsd/maven-4.0.0.xsd">
http://maven.apache.org/POM/4.0.0 http://maven.apache.org/xsd/maven-4.0.0.xsd">
http://maven.apache.org/POM/4.0.0 http://maven.apache.org/xsd/maven-4.0.0.xsd">
http://maven.apache.org/POM/4.0.0 http://maven.apache.org/xsd/maven-4.0.0.xsd">
http://maven.apache.org/xsd/maven-4.0.0.xsd</a>
                      <modelVersion>4.0.0</modelVersion>
                      <groupId>com.sparks.spark-streaming.basics
                      <artifactId>spark-streaming-basics</artifactId>
                      <version>0.0.1-SNAPSHOT
                              <spark.version>1.6.2</spark.version>
                      </properties>
  10
11
12
13
14
15
16
17
18
19
20
21
22
23
24
25
26
27
28
                      <dependencies>
                              <!-- https://mvnrepository.com/artifact/org.apache.spark/spark-streaming_2.11 -->
                              <dependency>
                                      <groupId>org.apache.spark</groupId>
                                      <artifactId>spark-streaming_2.11</artifactId>
<version>${spark.version}</version>
                              </dependency
                              <!-- https://mvnrepository.com/artifact/org.apache.spark/spark-streaming-twitter_2.11 -->
                              <dependency>
                                       <groupId>org.apache.spark</groupId>
                                      <artifactId>spark-streaming-twitter 2.11</artifactId>
<version>${spark.version}</version>
                              </dependency>
<dependency>
                                     <groupId>org.apache.opennlp</groupId>
<artifactId>opennlp-tools</artifactId>
                                      <version>1.5.3
                              </dependency>
  29
30
                              <dependency>
    <groupId>org.twitter4j</groupId>
  31
32
                                      <artifactId>twitter4j-core</artifactId>
<version>4.0.1</version>
  33
34
35
36
37
38
39
                              </dependency>
                              <dependency>
                                      <groupId>org.twitter4j</groupId>
                                      <artifactId>twitter4j-stream</artifactId>
<version>4.0.1</version>
                              </dependency>
                              <dependency>
  40
41
                     <groupId>org.slf4j</groupId>
<artifactId>slf4j-simple</artifactId>
  42
43
44
45
46
                     <version>1.7.7
                     </dependency>
                     <dependency>
                     <groupId>org.apache.opennlp</groupId>
<artifactId>opennlp-tools</artifactId>
                     <version>1.5.3
                      </dependency>
  49
                      </dependencies>
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Spark Twitter Data Processor.java

```
📑 Sentiment.Analysis WithCount.java 🗵 📙 OpenNLPCategorizer.java 🗵 🚆 Spark TwitterDataProcessor.java 🗵 📑 sparkit.java 🗵
      package goddmit;
      import org.apache.spark.SparkConf;
      import org.apache.spark.api.java.JavaPairRDD;
      import org.apache.spark.api.java.function.VoidFunction;
      import org.apache.spark.streaming.Duration;
      import org.apache.spark.streaming.api.java.JavaDStream;
      import org.apache.spark.streaming.api.java.JavaPairDStream;
      import org.apache.spark.streaming.api.java.JavaReceiverInputDStream;
      import org.apache.spark.streaming.api.java.JavaStreamingContext;
      import org.apache.spark.streaming.twitter.TwitterUtils;
 10
 12 import scala.Tuple2;
 13 import twitter4j.Status;
 14 import twitter4j.auth.Authorization;
     import twitter4j.auth.OAuthAuthorization;
     import twitter4j.conf.Configuration;
      import twitter4j.conf.ConfigurationBuilder;
 17
 18
 19
     import com.google.common.collect.Iterables;
 20
 21 | Dpublic class SparkTwitterDataProcessor {
 22 白
           public static void main(String[] args) {
 23
 24
               // Prepare the spark configuration by setting application name and master(only) node "local"
 25
               final SparkConf sparkConf = new SparkConf().setAppName("Twitter Data Processing").setMaster("local[10]");
 26
               // Create Streaming context using spark configuration and duration for which messages will be batched and fed to
 27
               Spark Core
 28
               final JavaStreamingContext streamingContext = new JavaStreamingContext(sparkConf, Duration.apply(10000));
 29
 30
               // Prepare configuration for Twitter authentication and authorization
 31
               final Configuration conf = new ConfigurationBuilder().setDebugEnabled(false)
 32
                                                 .setOAuthConsumerKey("
 33
                                                 .setOAuthConsumerSecret("
                                                 .setOAuthAccessToken("
 34
 35
                                                 .setOAuthAccessTokenSecret("
 36
                                                 .build();
Java source file
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```

```
Sentiment Analysis With Count, java 🗵 📙 Open NLP Categorizer, java 🗵 📙 Spark Twitter Data Processor, java 🗵 📙 sparkit, java 🗵
 38
               // Create Twitter authorization object by passing prepared configuration containing consumer and access keys and tokens ^
 39
               final Authorization twitterAuth = new OAuthAuthorization(conf);
 40
 41
               // Create a data stream using streaming context and Twitter authorization
 42
               final JavaReceiverInputDStream<Status> inputDStream = TwitterUtils.createStream(streamingContext, twitterAuth, new
               String[]{});
 43
 44
               // Create a new stream by filtering the non english tweets from earlier streams
 45
               final JavaDStream<Status> enTweetsDStream = inputDStream.filter((status) -> "en".equalsIgnoreCase(status.getLang()));
 46
 47
               // Convert stream to pair stream with key as user screen name and value as tweet text
 48
               final JavaPairDStream<String, String> userTweetsStream =
 49
                                        enTweetsDStream.mapToPair(
 50
                                             (status) -> new Tuple2<String, String>(status.getUser().getScreenName(), status.getText())
 51
                                        1:
 52
 53
               // Group the tweets for each user
 54
               final JavaPairDStream<String, Iterable<String>> tweetsReducedByUser = userTweetsStream.groupByKey();
 55
 56
               // Create a new pair stream by replacing iterable of tweets in older pair stream to number of tweets
 57
               final JavaPairDStream<String, Integer> tweetsMappedByUser = tweetsReducedByUser.mapToPair(
 58
                           userTweets -> new Tuple2<String, Integer>(userTweets. 1, Iterables.size(userTweets. 2))
 59
                       );
 60
 61
               // Iterate over the stream's RDDs and print each element on console
 62
               tweetsMappedByUser.foreachRDD((VoidFunction<JavaPairRDD<String, Integer>>)pairRDD -> {
 63
                   pairRDD.foreach(new VoidFunction<Tuple2<String,Integer>>() {
 64
 65
                       @Override
 66
                       public void call(Tuple2<String, Integer> t) throws Exception {
 67
                           System.out.println("#"+t. 1() + "," + t. 2());
 68
 69
 70
                   });
 71
               });
 72
 73
               // Triggers the start of processing
 74
               streamingContext.start();
 75
 76
               // Keeps the processing live by halting here unless terminated manually
 77
               streamingContext.awaitTermination();
 78
 79
 80
ava source file
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```

Sentiment Analysis with count.java

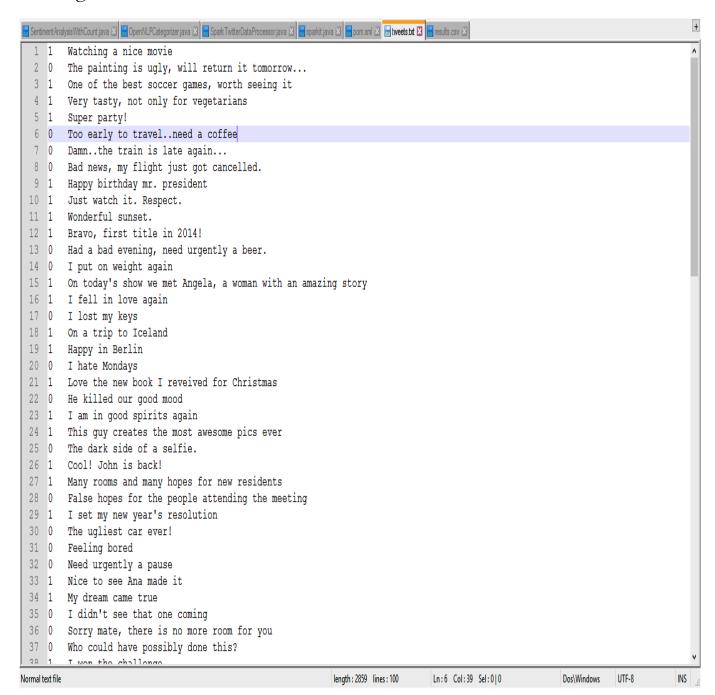
```
🗏 Sentiment Analysis With Count, java 🗵 📙 Open NLP Categorizer java 🗵 📙 Spark Twitter Data Processor java 🗵 📑 sparkit java 🗵
      package goddmit;
     import java.io.BufferedWriter;
  4 import java.io.FileInputStream;
  5 import java.io.FileWriter;
  6 import java.io.IOException;
  7 import java.io.InputStream;
    import java.util.Scanner;
 10 import opennlp.tools.doccat.DoccatModel;
 import opennlp.tools.doccat.DocumentCategorizerME;
 12 import opennlp.tools.doccat.DocumentSampleStream;
 13 import opennlp.tools.util.ObjectStream;
 14 import opennlp.tools.util.PlainTextByLineStream;
 15 import twitter4j.Query;
 16 import twitter4j.QueryResult;
 17 import twitter4j.Status;
 18 import twitter4j.Twitter;
 19 import twitter4j.TwitterException;
 20 import twitter4j.TwitterFactory;
 21 import twitter4j.conf.ConfigurationBuilder;
 23 Epublic class SentimentAnalysisWithCount {
 24
          DoccatModel model;
 25
          static int positive = 0;
 26
          static int negative = 0;
 27
          public static void main(String[] args) throws IOException, TwitterException {
 28 白
 29
              String line = "";
 30
 31
              //Constructor
 32
              SentimentAnalysisWithCount twitterCategorizer = new SentimentAnalysisWithCount();
 33
 34
              //Train Open NLP model
              twitterCategorizer.trainModel();
 36
                                                               Java source file
```

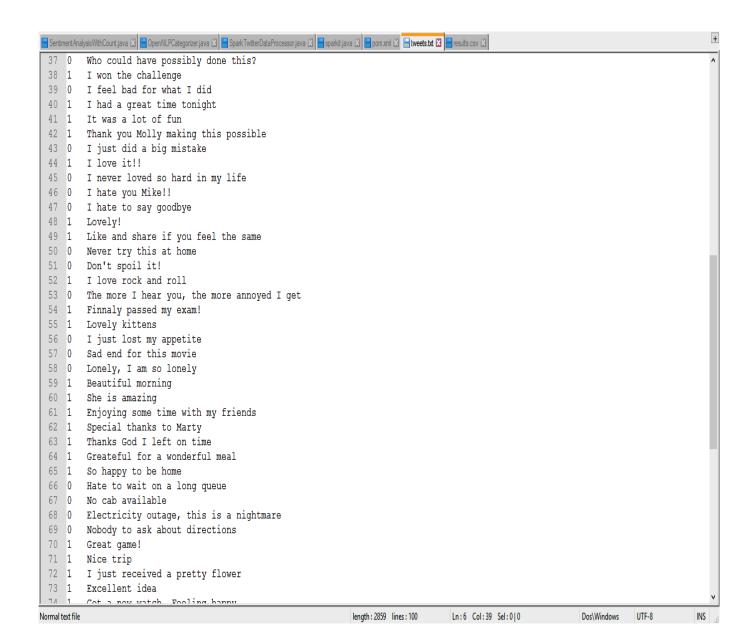
```
🔚 Sentiment Analysis With Count java 🛭 📙 Open NLP Categorizer java 🖸 📙 Spark Twitter Data Processor java 🗵 📑 sparkit java 🖸
 37
               // Twitter Keys and Access Token
 38
               ConfigurationBuilder cb = new ConfigurationBuilder();
 39
               cb.setDebugEnabled(true)
 40
                        .setOAuthConsumerKey("
 41
                        .setOAuthConsumerSecret("
 42
                        .setOAuthAccessToken("
 43
                        .setOAuthAccessTokenSecret("
                                                                                                       ");
 44
 45
               //Build the object
               TwitterFactory tf = new TwitterFactory(cb.build());
 46
 47
 48
               //Twitter instance is created for streaming tweets.
 49
               Twitter twitter = tf.getInstance();
 50
 51
               //Search for Topic
 52
               String search topic=new String();
 53
               System.out.println("Search: ");
 54
               Scanner me=new Scanner (System.in);
 55
               search topic=me.nextLine();
 56
 57
               //Search in twitter using Query
 58
               Query query = new Query(search topic);
 59
               QueryResult result = twitter.search(query);
 60
 61
               //Classify incoming tweets by reading tweet text from tweet data
 62
               int result1 = 0;
 63
               for (Status status : result.getTweets()) {
 64
                   result1 = twitterCategorizer.classifyNewTweet(status.getText());
 65
                   if (result1 == 1) {
 66
                        positive++;
 67
                   } else {
 68
                        negative++;
 69
 70
 71
```

```
🔚 Sentiment Analysis With Count java 🛽 📙 Open NLP Categorizer java 🔀 🔡 Spark Twitter Data Processor java 🗴 📑 sparkit java 🔀
 72
               //Write the count
 73
               BufferedWriter bw = new BufferedWriter(new FileWriter(" results.txt"));
 74
               bw.write("Positive Tweets," + positive);
 75
               bw.newLine();
 76
               bw.write("Negative Tweets," + negative);
 77
               bw.close();
 78
               me.close();
 79
 80
 81
           public void trainModel() {
 82
               InputStream dataIn = null;
 83
 84
               //Read training data from txt
 85
               try {
 86
                   dataIn = new FileInputStream(" tweets.txt");
 87
                   ObjectStream lineStream = new PlainTextByLineStream(dataIn, "UTF-8");
 88
                   ObjectStream sampleStream = new DocumentSampleStream(lineStream);
 89
 90
                   // Specifies the minimum number of times a feature must be seen
 91
                   int cutoff = 3;
 92
                   int trainingIterations = 100;
 93
                   model = DocumentCategorizerME.train("en", sampleStream, cutoff,
 94
                            trainingIterations);
 95
               } catch (IOException e) {
 96
                   e.printStackTrace();
 97
               } finally {
 98
                   if (dataIn != null) {
 99
                       try {
100
                            dataIn.close();
101
                        } catch (IOException e) {
102
                            e.printStackTrace();
103
104
105
106
107
```

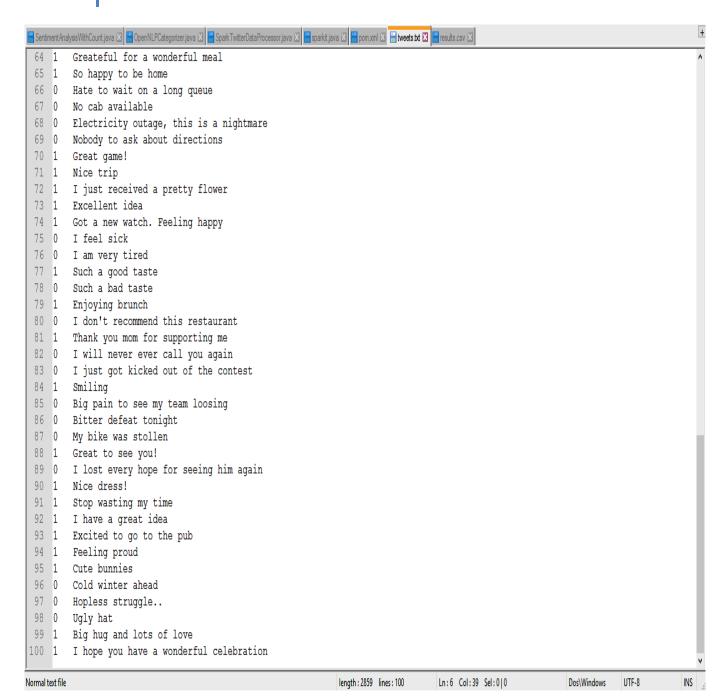
```
🖺 Sentiment Analysis With Count, java 🗓 📙 Open NLP Categorizer java 🖸 📙 Spark Twitter Data Processor java 🗵 📙 sparkit, java 🗵
108
           public int classifyNewTweet(String tweet) throws IOException {
109
               DocumentCategorizerME myCategorizer = new DocumentCategorizerME (model);
110
               double[] outcomes = myCategorizer.categorize(tweet);
111
               String category = myCategorizer.getBestCategory(outcomes);
112
113
               //Use Open NLP lib to categorize
               System.out.print("-----\nTWEET :" + tweet + " ===> ");
114
               if (category.equalsIgnoreCase("1")) {
115
116
                   System.out.println(" POSITIVE ");
117
                   return 1:
118
               } else {
119
                   System.out.println(" NEGATIVE ");
120
                   return 0;
121
122
123
124 L}
lava source file
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```

Training Data: tweets.txt



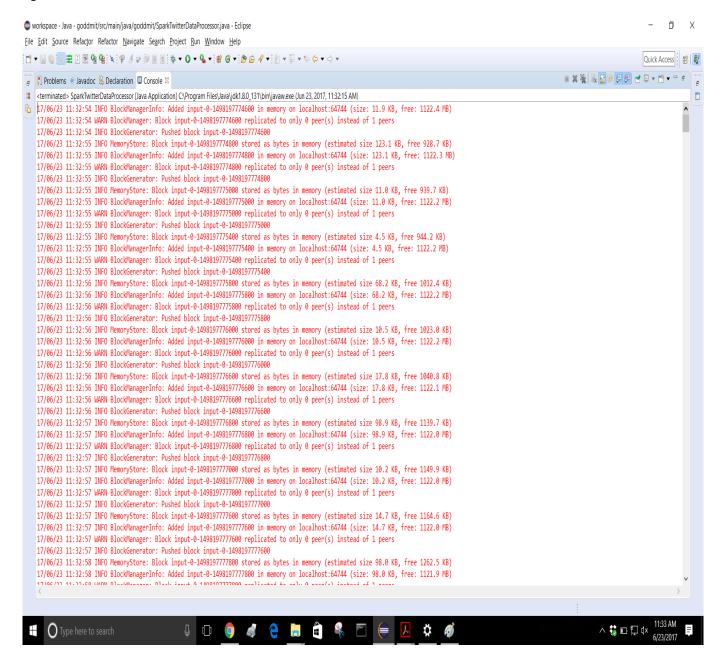


INDUSTRIAL TRAINING – II

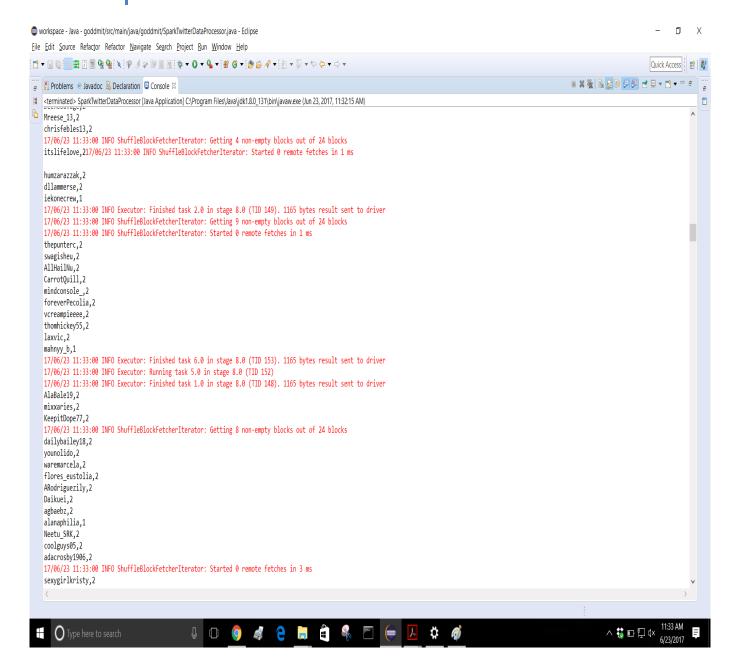


RESULTS:

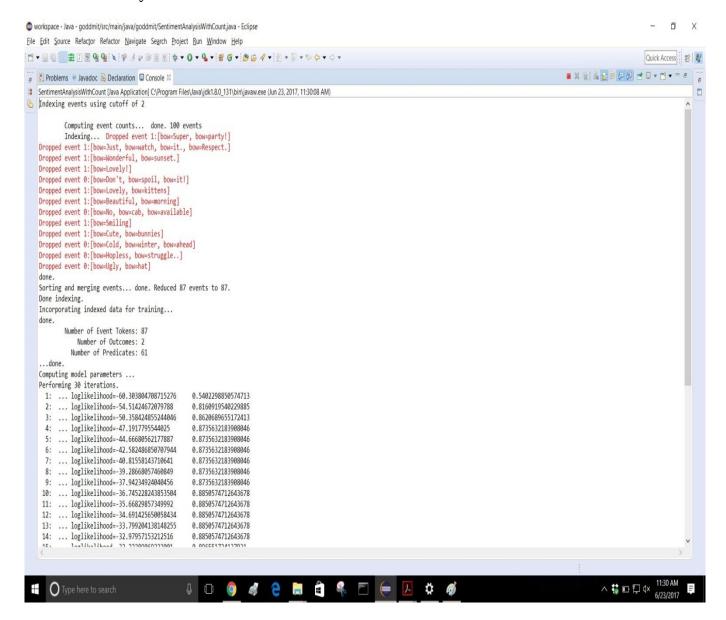
SparkTwitterDataProcessor:



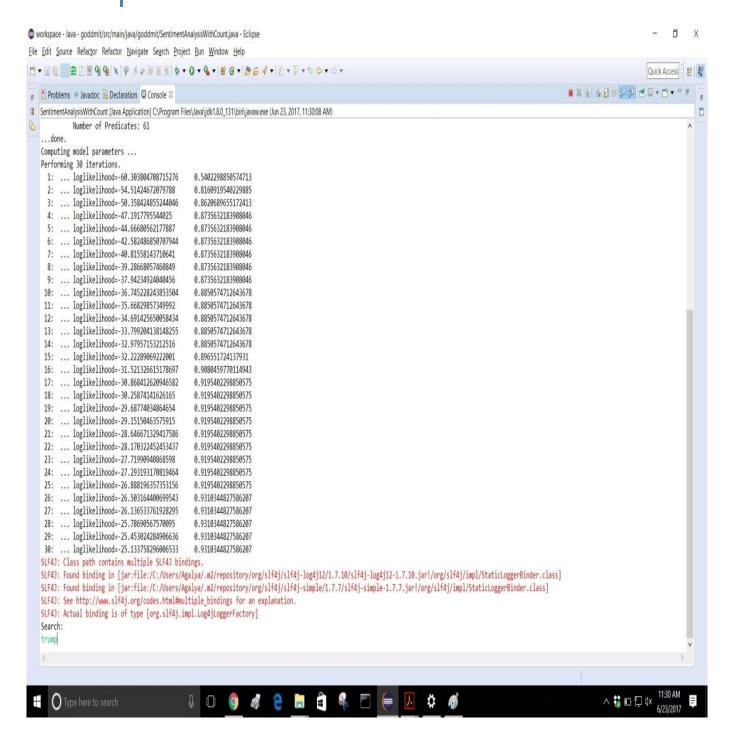
INDUSTRIAL TRAINING – II

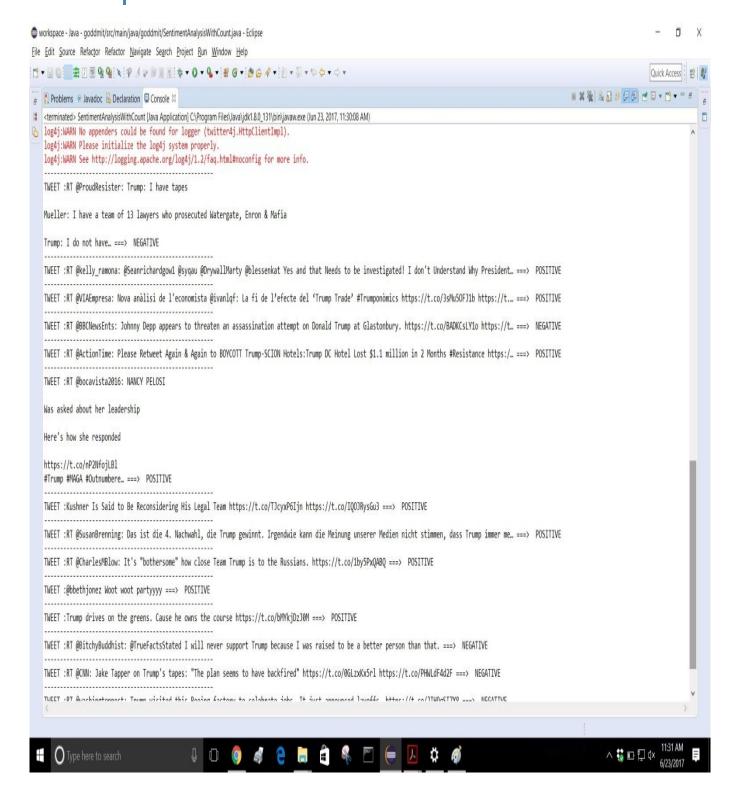


Sentiment Analysis with count:

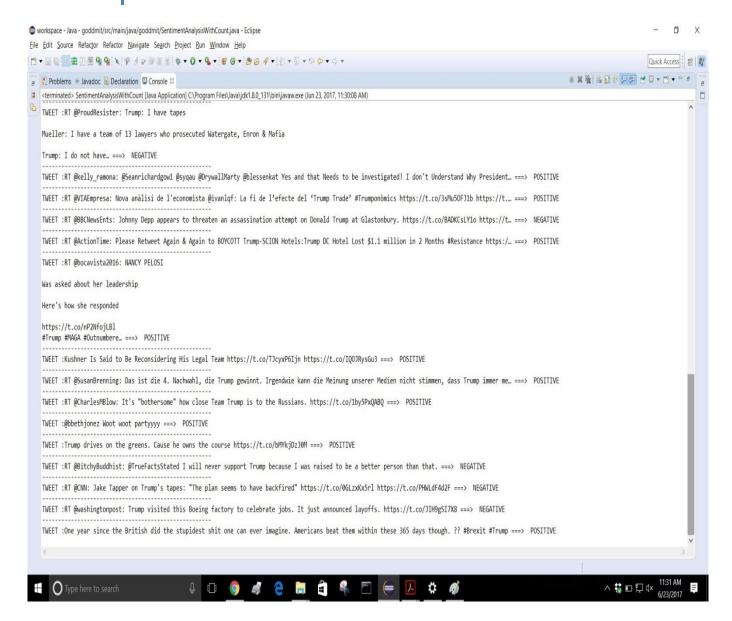


INDUSTRIAL TRAINING - II





INDUSTRIAL TRAINING - II



FUTURE SCOPES

This application can be easily implemented under various situations. We can add new features as and when we require. Reusability is possible as and when require in this application. There is flexibility in all the modules.

SOFTWARE SCOPE:

- *Extensibility*: This software is extendable in ways that its original developers may not expect. The following principles enhance extensibility like updating data locally while being offline and then syncing the server when there is internet connection. We can also delete values either offline or online and then update it as and when there is an internet connection.
- Reusability: Reusability is possible as and when require in this application.
- *Understandability:* A method is understandable if someone other than the creator of the method can understand the code (as well as the creator after a time lapse). We use the method, which small and coherent helps to accomplish this.
- *Cost-effectiveness*: Its cost is under the budget and make within given time period. It is desirable to aim for a system with a minimum cost subject to the condition that it must satisfy the entire requirement. Scope of this document is to put down the requirements, clearly identifying the information needed by the user, the source of the information and outputs expected from the system.

REFERENCES:

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http://www.infoq.com/articles/apache-sparkstreaming?utm_source=apachesparkseries&utm_medium=link&utm_campaign=internal

vi. Spark Tutorial by Edureka

https://www.edureka.co/blog/sparkstreaming/?utm_source=blog&utm_medium=left-menu&utm_campaign=sparktutorial

vii. Tutorialspoint

https://www.tutorialspoint.com/apache_spark/apache_spark_quick_guide.htm