A PROJECT REPORT

ON

"TWITTER TREND ANALYSIS"

SUBMITTED IN PARTIAL FULFILLMENT OF THE REQUIREMENT FOR THE DEGREE OF BACHELOR OF COMPUTER ENGINEERING

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CERTIFICATE

This is to certify that the project entitled "TWITTER TREND ANALYSIS" is bona fide work of SANKALP NILEKAR, SUDEEP RAWAT and RAHUL VERMA submitted to the University of Mumbai in partial fulfillment of the requirement for the award of the degree of Undergraduate in Computer Engineering.

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PROJECT APPROVAL

This project report entitled "TWITTER TREND ANALYSIS" by

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degree of Computer Engineering.

Internal Examiner	External Examiner
Date:	Date:
Place·	Place.

DECLARATION

We declare that, this written submission represents our ideas in our own words and where others' ideas or words have been included; we have adequately cited and referenced the original sources. We also declare that we have adhered to all principles of academic honesty and integrity and have not misrepresented or fabricated or falsified any idea/data/fact/source in our submission. We understand that any violation of the above will be cause for disciplinary action by the Institute and can also evoke penal action from the sources which have thus not been properly cited or from whom proper permission has not been taken when needed.

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ABSTRACT

The community of users participating in social media tends to share common interests at the same time, giving rise to what are known as social trends. A social trend reflects the voice of a large number of users which, for some reason, becomes popular in a specific moment. Through social trends, users, therefore, suggest that some occurrence of wide interest is taking place and subsequently triggering the trend. In this work, we explore the types of triggers that spark trends on the microblogging site Twitter and introduce a typology that includes the following four types: news, ongoing events, memes, and commemoratives. The user will be allowed to search for the latest trends by inputting a keyword into the search field. Based on user-provided keyword, the system will search for similar keywords in database and summarize the total count to provide the trending tweets on twitter. The trending tweets with the hashtag (#) will be displayed first and then the rest words will be displayed. By clicking on every trending tweet, the user commented tweets will be displayed. User can view all the tweets from the searched keyword. One of the main features on the homepage of Twitter shows a list of top terms so-called trending topics at all times. These terms reflect the topics that are being discussed most at the very moment on the site's fastflowing stream of tweets. In order to avoid topics that are popular regularly (e.g., good morning or goodnight on certain times of the day), Twitter focuses on topics that are being discussed much more than usual, i.e., topics that recently suffered an increase of use, so that it trended for some reason.

Trending topics have attracted big interest not only among the users themselves but also among other information consumers such as journalists, real-time application developers, and social media researchers. Being able to know the top conversations being discussed at a given time helps keep updated about current affairs, and discover the main concerns of the community. Twitter defines trending topics as "topics that are immediately popular, rather than topics that have been popular for a while or on a daily basis". However, no further evidence is known about the algorithm that extracts trending topics. It is assumed that the list is made up by terms that appear more frequently in the most recent stream of tweets than the usual expected.

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INTRODUCTION

1.1 Aim and Objective:

In the times of information age, the magnitude of online social media activity has reached an unprecedented level. Hundreds of millions of users spend hours online every day to stay connected and communicate with the rest of the world. Millions of users participate in these social networks of Social awareness streams. [19] People generate huge amount of data every day on various social media networks, which in aggregate indicate the interests and current attention of the local and global communities. There are many events and topics discussed on Twitter. Some topics may get a lot of attention and some may not. Some of these topics become very popular and focus of interests for a large number of people. The connections and the nature of social network let information disseminate to a large number of other people, a phenomenon known as going "viral". These popular topics of discussions are also called "trends" in the social network. These trends are very dynamic and temporal in nature which exposes the expose the aggregate interests and attention of global and local communities.

Trends in social networks are of high significance and a major point of interest in both the industry [19] and the research community [8, 15]. Many applications on web and business can be immensely benefitted from knowing what is currently "trending", which represents an answer to the age-old query what are people talking about [9]. From stock exchange making real-time decision to search engines delivering more updated, relevant search results. Twitter is one of the most popular social networking and micro-blogging service, which had more than 200 Million registered users by 2013, producing 400 Million tweets every day [17]. As a microblogging website it allows its users to create a short text message of 140 characters as their posts called "tweets". There are also many different ways for users to update their tweets, including the mobile phone, web and text messaging tools [14] and so on. Twitter is also very real-time in nature. In pasts, several events were reported on twitter as news hours

earlier than the mainstream media. [10]. Hence twitter is a very robust source for getting the real-time trends in the web.

The numbers of active users and tweets generated daily are enormous and hence, they collectively can give crucial clues to several interesting problems such as public opinion analysis and hot trend detection. Twitter employs a social model called following [14], in which the user is allowed to choose any other users that they want to follow without any permission or reciprocating by following them back. The one they follow is their friend, and they are the follower. Being a follower on Twitter means that they receive all the updates of their friends [4]. This makes twitter a directed social network where directed links could represent anything from intimate friendships to common interests, or even a passion for breaking news or celebrity gossip. Such directed links determine the flow of information and hence indicate a user's influence on others a concept that is crucial in sociology and viral marketing. The major drawback with using Twitter as a source of information is that not all of the tweets are informative. Contrary to it, the majority of tweets are "chatty" or "spammy" in nature [7]. So, it's crucial to filter out this noise from data to use the useful "informative" tweets. Hence, we need a system, which could separate useful tweets, which essentially means a classifier model to classify "informative" tweets from "chat" tweets. This system should work in a single pass and also be robust and fast enough to process up to 400 million tweets a day. Our attempt in this thesis is to provide such a model of a framework which given topic wise tweets clusters will be able to detect current trends and also predict some upcoming trends in their early stage using the social graph of twitter.

1.2 Problem Statement:

Our problem statement can be formally defined as follows.

Create an automated system, which takes in a continuous stream of raw tweets, processes these tweets to filter out "noise" and get relevant informative tweets. Further mine these "filtered" tweets to detect and predict evolving "trending" topics in their very early stage.

There are two key aspects to this problem:

- Creating a model to process real-time tweets feed to filter out the "noise" and output only relevant tweets.
- Creating a system to store the social graph of twitter users such that it can be
 efficiently used as a data structure for answering various user queries and get
 relations for a given user.

1.3 Scope of the Project:

• Twitter Trend Analysis helps us to know what people are talking about:

Twitter Trend analysis searches for the trends happening around the world and thus we can know about the things people are talking about or the things people are currently interested in talking. What we do here in Twitter trend analysis is filter out the noise in the tweets and provide the user with the relevant information. Using this we come to know what the users require and such data can be used by various companies to provide specific services to customers. This data provided by Twitter trend analysis is very helpful for activity-based advertisements which are helpful for both the company providing it and the user as well.

• We can explore the trending topics using Trend Analysis:

Twitter helps us to provide information instantaneously and thus it can be used to spread important information around the world without any delays and maximum coverage. News channels prefer twitter for giving any information due to its reach. Popular topics in the region can be found using Twitter Trend Analysis. This helps with getting people aware of the things that are currently trending in the region.

• It helps us reach a wider audience:

Twitter has a large user base, which could include your potential customers. Using hashtags can help you reach an audience interested in a particular topic or in a particular location. The platform allows direct two-way communication with your customers. Because it's a business interaction, if you do it well it shows your business in a positive light.

• Twitter Trend Analysis will help us remove spam:

Twitter is filled with spam messages or noise just like any other social media platform which introduces lots of problems in the work of many people. Twitter Trend Analysis helps us solve all these issues by filtering out the noise and displaying just the relevant information that is needed by the user. This helps in increased productivity and reduces the work of the users. Reducing spam also reduces duplication and virus attacks. The noise removal taking place in Twitter Trend Analysis is really important and is one the most important aspect of it.

• Future Tweets Will Be Searched on Search Engines:

If you have searched on Google, you must have seen that tweets appear among the search results. In the future twitter, we expect that searching tweets will apply on Google and Bing. This means that you will be able to search for a specific tweet. This will be achieved by entering the tweet wording on the search bar. The relating tweets will appear as tweets results on your browser. Currently, the tweets from big brands appear on Google searches. Google featured tweets will be within everyone's grasp soon.

LITERATURE REVIEW

Trend analysis and based on that predicting public opinions. It plays important role, many researchers working on the automatic technique of extraction and analysis of huge amount of twitter data. In [1] the author compares six trend detection method and find that standard natural language processing technique perform well for social streams on a particular topic. They conclude that n-gram gives the best performance other than state-of-art techniques. In [4], the authors have used three different machine learning algorithms Naïve Bayes, Decision Trees and Support Vector Machine for sentiment classification of Arabic dataset which was obtained from Twitter. This research has followed a framework for Arabic tweets classification in which two special sub-tasks were performed in preprocessing, Term Frequency-Inverse Document Frequency (TF-IDF) and Arabic stemming. They have used one dataset with three algorithms and performance has been evaluated on the basis of three different information retrieval metrics precision, recall, and f-measure. In [6] author proposed supervised learning techniques to classify twitter trending topic for that they use text-based and network-based classifier and conclude C5.0 gave the best performance.

In [19] author propose model which predict public opinion on the political event by Applying different classifier which predicts whether the mood is positive or negative. In [26], the authors proposed a way to get the pre-labelled data from twitter which can be used to train the SVM classifier. They used the twitter hashtags to judge the polarity of a tweet. To analyze the accuracy of the proposed technique, a test study on the classifier was conducted which showed the result with the accuracy of 85%. The authors in [27] introduced a new technique to classify the sentiment of tweets as positive or negative. They presented and discussed the results of machine learning algorithms for twitter sentiment analysis by using distant supervision. Training data, the authors used consisted of tweets with emotions which were used as noisy labels. According to authors, the machine learning algorithms such as Naive Bayes, Maximum Entropy and SVM when trained with emotion tweets can have accuracy more than 80%. The study also highlighted the steps used in the preprocessing stage of classification for high accuracy. In [28] sentiment analysis performs using SVM in that two pre-classified datasets of tweets are used then do a comparative analysis, they use measures Precision, Recall and F-Measure.

Title of Paper	Author	Published in	Results
Sensing	Luca Maria Aiello,	IEEE	Author compare six trend
Trending Topics	Georgios Petkos,	Transactions on	detection method and find
in Twitter	Carlos Martin,	Multimedia, Vol.	that standard natural language
	David Corney,	15, No. 6, October	processing technique perform
	Symeon	2013.	well for social streams on
	Papadopoulos,		particular topic. They
	Ryan Skraba, Ayse		conclude that n-gram give
	Göker, Ioannis		best performance other than
	Kompatsiaris		state of art techniques
Comparison of	Altawaier, M. M., &	International	The Authors have used one
Machine	Tiun.	Journal on	dataset with three algorithms
Learning		Advanced	and performance has been
Approaches on		Science,	evaluated on the basis three
Twitter Trend		Engineering and	different information retrieval
Analysis		Information	metrics precision, recall, and
		Technology, 6(6)	f-measure.
Twitter	Kathy Lee, Diana	2011 11th IEEE	Author proposed supervised
Trending Topic	Palsetia,	International	learning techniques to
Classification	Ramanathan	Conference on	classify twitter trending topic
	Narayanan, Md.	Data Mining	for that they use text based
	Mostofa Ali		and network-based classifier
	Patwary, Ankit		and conclude C5.0 gave best
	Agrawal,		performance
Study using	Zgheib, W. A., &	CS224N Project	The authors proposed a way
Support Vector	Barbar	Report, Stanford,	to get the pre labeled data
Machines to		1(2009), 12	from twitter which can be
Classify the			used to train SVM classifier.
Sentiments of			They used the twitter hash
Tweets			tags to judge the polarity of
			tweet. To analyze the
			accuracy of proposed
			technique

REQUIREMENT ANALYSIS

• Tweet impressions:

Under the Tweets section, you can find a list of all your Tweets and the number of impressions. You can see individual Tweet performance, as well as recent months or a 28-day overview of cumulative impressions. Capitalize on this information by repurposing Tweets that gained the most impressions, or creating Tweets on a similar subject. You can also use the cumulative overview to compare monthly activity.

• Tweet engagements and engagement rate:

Similar to impressions, the Tweets section also shows your Tweets engagement, or the number of interactions your Tweet has received, as well as the engagement rate, which is engagements divided by impressions. If your Tweets are receiving little engagement, you may want to rethink your subject matter and format, for instance, you may want to add photo or video to your content mix, which tends to generate more engagement.

• Event and trending topic data:

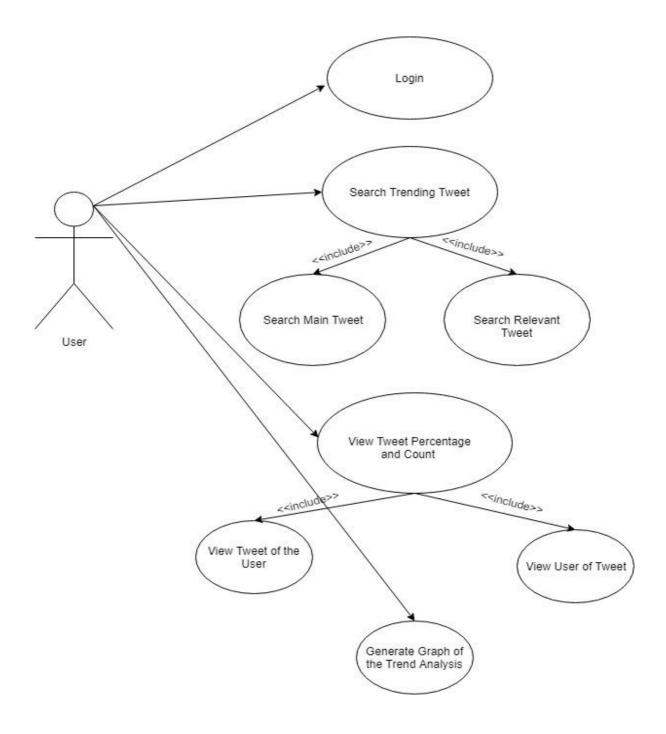
Discover upcoming holidays, events, and recurring trends, and find out who's Tweeting about them. This is great way to find potential new content ideas, and conversations to join in on.

• Video content performance:

If you're using video as part of your content strategy, you can track your video views, as well see a bigger picture of how people are responding to your videos. For instance, are they watching it to completion? If you want to fine-tune your Twitter strategy, spending some time understanding your Twitter analytics is a great place to start. Get started by viewing your Twitter analytics dashboard today.

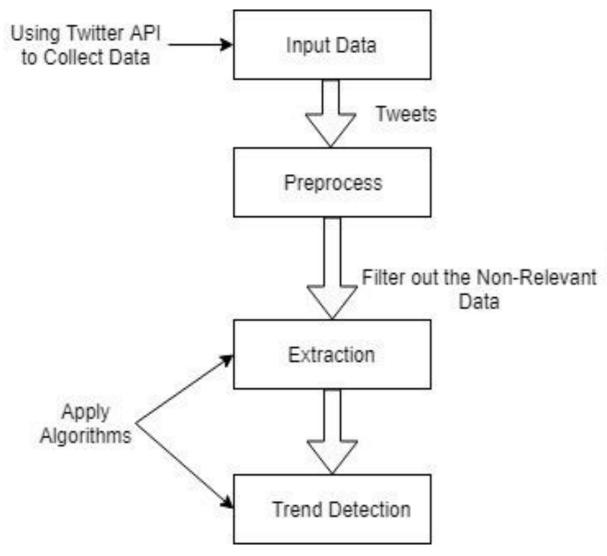
3.1 Use Case Diagram:

A Use Case Diagram is a graphic depiction of the interactions among the elements of a system. A Use Case Diagram contains four components- Boundary, Actors, use cases and Relationships among actors and use cases.



Use Case Diagram for Twitter Trend Analysis (fig. 3.1.1)

3.2 Proposed Model:



Model for Twitter Trend Analysis (fig. 3.2.1)

The model has following steps:

- Data collection of tweets
- Pre-process tweet
- Extraction
- Trend Detection

1) Dataset: Collect tweet data through twitter streaming API. Which download tweets in JSON format. We can apply keyword, hashtag, username to download tweets related to them.

2) Pre-processing: Tweet pre-processing module having several stages. After downloading tweets, we have to extract text data form that and discard video, audio, image etc. store English text which is retrieve form tweet. Then remove @, #, URL and other punctuation form tweets and apply stop word remove, word tokenize.

3) Extraction: After pre-processing stage next module is Extraction which is done in two way through Term frequency calculation and pos tagging. We can view the Count, Strength etc. after the extraction of the data, which can therefore, be used to view tweets using similar hashtags from different users

4) Trend Detection: We can determine trend by using TF-IDF calculation. And predict positive, negative, neutral mood tendency by applying machine learning algorithms. Show the graph for the given input by the user.

3.3 Process:

The system comprises of 4 major modules with its description as follows:

1) Login:

User need to login first using valid credentials to access the system.

2) Search for Latest Trends:

After successful login, user can search for latest trending tweets by entering the keyword in the search column.

3) View Latest Trending Tweet:

Based on user-inputted keyword, the search results will be displayed in form of trending tweets.

4) View Tweets:

User can click on respective trending tweet to view the message twitted by other users

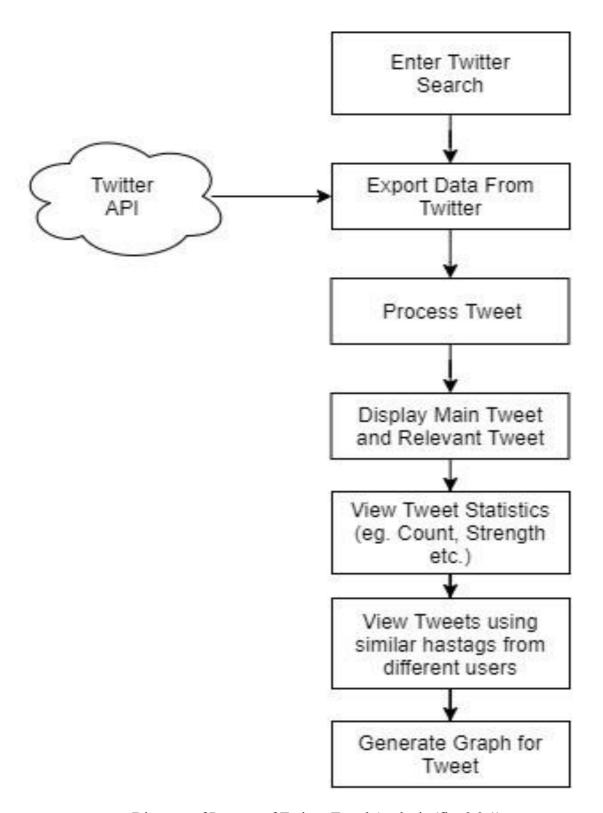


Diagram of Process of Twitter Trend Analysis (fig. 3.3.1)

3.4 Software and Hardware Requirements:

Software Requirements:

- Windows 7 or higher.
- Python 3.0 or higher.

Hardware Components:

- Processor i3
- Hard Disk 5 GB
- Memory 1GB RAM
- Internet Connection

PROJECT DESIGN

4.1 Collecting data via Twitter API:

As the mentioned earlier core of our approach is the social graph, which consists of twitter users and relations among them. The current size of twitter has more than 200 million nodes or users in it. [17] Twitter exposes these relations to the public via their API. [4] where relations like "followers" and "friends" (users a person follows) can be collected via API calls. The major blocking factor here is the rate limit to a number of calls, which can be made to API. We created more than 10 different API keys for created several instances of nodes making calls to Twitter API. We wrote a python library to communicate with the API. Also, the Twitter API was upgraded several times by Twitter resulting in a change of API methods and responses. Hence required changes in our code accordingly. Also due to heavy traffic or technical glitches, there were frequent outages in the API service causing further delay and blocking.

4.2 Tweets clusters:

We need all the tweets related to a particular topic together which we call a cluster. This tweets cluster tells us who all users are involved in a discussion of that particular topic.

4.3 Twitter API:

When someone wants to access Twitters APIs, they are required to register an application. By default, applications can only access public information on Twitter. Certain endpoints, such as those responsible for sending or receiving Direct Messages, require additional permissions from us before they can access your information. These permissions are not granted by default, we choose on a per-application basis whether to provide this access, and can control all the applications authorized on our account.

The Twitter APIs include a wide range of endpoints, which fall into five primary groups:

• Accounts and users:

Twitter allows developers to programmatically manage an account's profile and settings, mute or block users, manage users and followers, request information about an authorized account's activity, and more. These endpoints can help citizen services like the Commonwealth of Virginia's Department of Emergency Management which provides information to residents about emergency responses and emergency alerts.

• Tweets and replies:

Twitter makes public Tweets and replies available to developers, and allow developers to post Tweets via our API. Developers can access Tweets by searching for specific keywords, or requesting a sample of Tweets from specific accounts. These endpoints are used by NGOs like the UN to identify, understand and counter misinformation around public health initiatives. For example, in Indonesia, there were persistent rumors that vaccinations either contained pork product or caused infertility. Understanding how the rumors started and were spread allowed the UN to get a team on the ground to help dispel these myths, which caused particular concern in this Muslim-majority nation. Similarly, Twitter helps researchers listen for early symptomatic indications of disease outbreaks and monitor their spread. A team at Northeastern University recently developed a new flu-tracking technique using Twitter data that forecasts outbreaks up to six weeks in advance which is much earlier than many other models, without sacrificing accuracy.

Direct Messages:

Twitters Direct Message endpoints provide access to the DM conversations of users who have explicitly granted permission to a specific application. Twitters DM APIs provide limited access to developers to create personalized experiences on Twitter, like Wendy's March Madness bracket builder. For accounts they own or manage, businesses can create these human- or chatbot-powered conversational experiences to communicate directly with customers for customer service, marketing, and brand engagement experiences.

• Ads:

Twitter provides a suite of APIs to let developers, like Sprinklr, help businesses automatically create and manage ad campaigns on Twitter. Developers can use public

Tweets to identify topics and interests, and provide businesses with tools for running advertising campaigns to reach the diverse audiences on Twitter.

• Publisher tools and SDKs:

Twitter provides tools for software developers and publishers to embed Twitter timelines, share buttons, and other Twitter content on webpages. These tools allow brands to bring live, public conversations from Twitter into their web experience and make it easy for their customers to share information and articles from their sites.

DESIGN IMPLEMENTATION

5.1 GUI:

A GUI (graphical user interface) is a system of interactive visual components for computer software. A GUI displays objects that convey information, and represent actions that can be taken by the user. The objects change color, size, or visibility when the user interacts with them. GUI objects include icons, cursors, and buttons. These graphical elements are sometimes enhanced with sounds, or visual effects like transparency and drop shadows. A GUI is considered to be more userfriendly than a text-based command-line interface, such as MS-DOS, or the shell of Unix like operating systems. The GUI was first developed at Xerox PARC by Alan Kay, Douglas Engelbart, and a group of other researchers in 1981. Later, Apple introduced the Lisa computer with a GUI on January 19, 1983. A GUI uses windows, icons, and menus to carry out commands, such as opening, deleting, and moving files. Although a GUI operating system is primarily navigated using a mouse, a keyboard can also be used via keyboard shortcuts or the arrow keys. As an example, if you wanted to open a program on a GUI system, you would move the mouse pointer to the program's icon and double-click it. Unlike a command-line operating system or CUI, like Unix or MS-DOS, GUI operating systems are much easier to learn and use because commands do not need to be memorized. Additionally, users do not need to know any programming languages. Because of their ease of use and more modern appearance, GUI operating systems have come to dominate today's market. A pointing device, such as the mouse, is used to interact with nearly all aspects of the GUI. More modern (and mobile) devices also utilize a touch screen. However, as stated in previous sections, it is also possible to navigate a GUI using a keyboard.

5.2 Chart:

Charts are often used to ease understanding of large quantities of data and the relationships between parts of the data. Charts can usually be read more quickly than the raw data. They are used in a wide variety of fields, and can be created by hand (often on graph paper) or by computer using a charting application. Certain types of charts are more useful for presenting a given data set than others. For example, data that presents percentages in different groups (such as "satisfied, not satisfied, unsure") are often displayed in a pie chart, but may be more easily understood when presented in a horizontal bar chart. [2] On the other hand, data that represents numbers that change over a period of time (such as "annual revenue from 1990 to 2000") might be best shown as a line chart:

- Histogram: A histogram is an approximate representation of the distribution of numerical or categorical data. It was first introduced by Karl Pearson. To construct a histogram, the first step is to "bin" (or "bucket") the range of values—that is, divide the entire range of values into a series of intervals—and then count how many values fall into each interval. The bins are usually specified as consecutive, non-overlapping intervals of a variable. The bins (intervals) must be adjacent, and are often (but not required to be) of equal size. If the bins are of equal size, a rectangle is erected over the bin with height proportional to the frequency—the number of cases in each bin. A histogram may also be normalized to display "relative" frequencies. It then shows the proportion of cases that fall into each of several categories, with the sum of the heights equaling 1.
- Scatter Plot: A scatter plot can be used either when one continuous variable that is under the control of the experimenter and the other depends on it or when both continuous variables are independent. If a parameter exists that is systematically incremented and/or decremented by the other, it is called the control parameter or independent variable and is customarily plotted along the horizontal axis. The measured or dependent variable is customarily plotted along the vertical axis. A scatter plot can be used either when one continuous variable that is under the control of the experimenter and the other depends on it or when both continuous variables are independent.

TECHNOLOGIES USED

6.1 Python:

Python is an interpreted, high-level, general-purpose programming language. Created by Guido van Rossum and first released in 1991, Python's design philosophy emphasizes code readability with its notable use of significant whitespace. Its language constructs and object-oriented approach aim to help programmers write clear, logical code for small and large-scale projects. Python is dynamically typed and garbage-collected. It supports multiple programming paradigms, including structured (particularly, procedural), object-oriented, and functional programming. Python is often described as a "batteries included" language due to its comprehensive standard library. Python was conceived in the late 1980s as a successor to the ABC language. Python 2.0, released in 2000, introduced features like list comprehensions and a garbage collection system capable of collecting reference cycles. Python 3.0, released in 2008, was a major revision of the language that is not completely backward-compatible, and much Python 2 code does not run unmodified on Python 3. The Python 2 language, i.e. Python 2.7.x, was officially discontinued on 1 January 2020 (first planned for 2015) after which security patches and other improvements will not be released for it. With Python 2's end-oflife, only Python 3.5.x and later are supported. Python interpreters are available for many operating systems. A global community of programmers develops and maintains CPython, an open source reference implementation. A non-profit organization, the Python Software Foundation, manages and directs resources for Python and CPython development.

6.2 API:

An application programming interface (API) is a computing interface to a software component or a system, that defines how other components or systems can use it. It defines the kinds of calls or requests that can be made, how to make them, the data formats that should be used, the conventions to follow, etc. It can also provide extension mechanisms so that users can extend existing functionality in various ways and to varying degrees. An API can be entirely custom, specific to a component, or it can be designed based on an industry standard to ensure

interoperability. Some APIs have to be documented, others are designed so that they can be "interrogated" to determine supported functionality. Since other components/systems rely only on the API, the system that provides the API can (ideally) change its internal details "behind" that API without affecting its users. Today, with the rise of REST and web services over HTTP, the term is often assumed to refer to APIs of such services when given no other context (see the Web APIs section). Sometimes the term API is, by extension, used to refer to the subset of software entities (code, subcomponents, modules, etc.) that serve to actually implement the API of some encompassing component or system. In building applications, an API (application programming interface) simplifies programming by abstracting the underlying implementation and only exposing objects or actions the developer needs. While a graphical interface for an email client might provide a user with a button that performs all the steps for fetching and highlighting new emails, an API for file input/output might give the developer a function that copies a file from one location to another without requiring that the developer understand the file system operations occurring behind the scenes.

TESTING

The basic goal of software development process is to produce the software that has very few or no errors. In an effort to detect errors soon after they are introduced, each phase ends with verification activity such as review. However, most of these verification activities in the early phase of software development are based on human evaluation. So, it is impossible to detect all errors in that phase itself. Testing plays an important role in quality assurance of the software. It is dynamic method for verification and validation, where system to be tested, is executed under various constraints and its behavior or response is observed.

• Test Plans:

A test plan is a general document for the entire project that defines entire scope approach to be taken and schedule of testing as well as identifies thee test item for testing. The test planning can be done well before the actual testing commences and also can be done in parallel in design and implementation phase.

• Unit Testing:

Unit tests are tests that exercise a class or module in isolation from rest of the application, and form a significant element of Test Driven Development where failing tests are written to allow the application to be 'debugged into existence'. Here, various modules have been tested individually. This has been done manually to test cases with the help of which the application has been tested.

• Performance Testing:

Performance Testing has been done to measure the responsiveness of application to workload such as increased user request and the type of internet connection. The parameters were chosen randomly for anticipation of errors or bug which resolved until the application performed consistently.

• Compatibility Testing:

Compatibility Testing is a type of software testing used to ensure compatibility of the system application with various other hardware specifications. It is a type of nonfunctional testing. This type of testing helps us to find out how well the system application performs in a particular environment that includes hardware, network, operating system and other software. It is basically the testing of application or the product built with computing environment. It tests whether the software product that is built is compatible with other hardware, operating tests whether the software product that is built is compatible with other hardware, operating system, database or other system software or not. Twitter Trend Analysis is compatible with Windows 7 or later.

• System Testing:

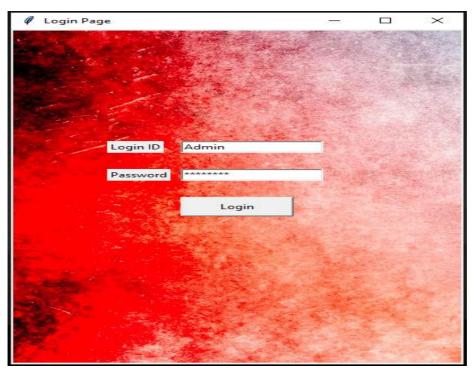
System Testing is actually a series of different tests whose sole purpose is to exercise the full computer based system. System testing falls under the black box testing category of software testing. White box testing is the testing of internal workings or coding implementations of software application. In contrast, black box or system testing is the opposite. System Testing involves the external working of the application from user's perspective. Here, after testing each module individually, all the classes and/or modules are integrated and tested as a whole application.

Test Cases:

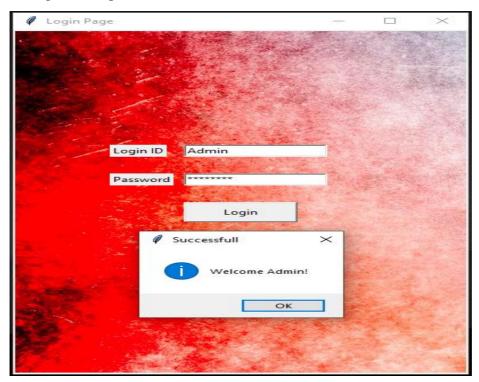
Sr. No.	Test Case Description	Expected Result	Actual Result
1.	User Login with correct user	Successful login	Pass
	id and password		
2.	For valid input	Output Screen (Username,	Pass
		Location, Tweets, Date) should	
		be displayed	
3.	For invalid input	Display "Please enter a valid	Pass
		Hashtag''	

RESULTS

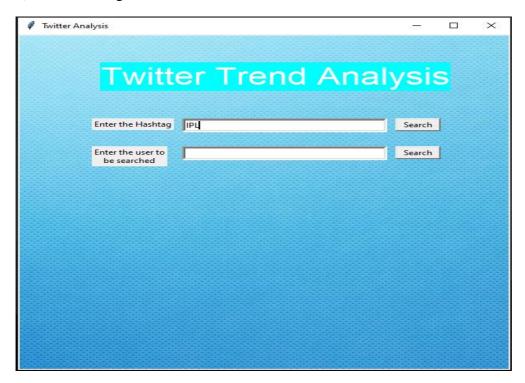
1) Login Page:



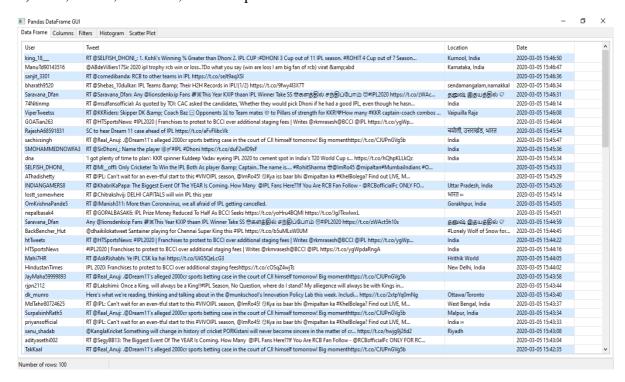
2) Login Attempt Successful:



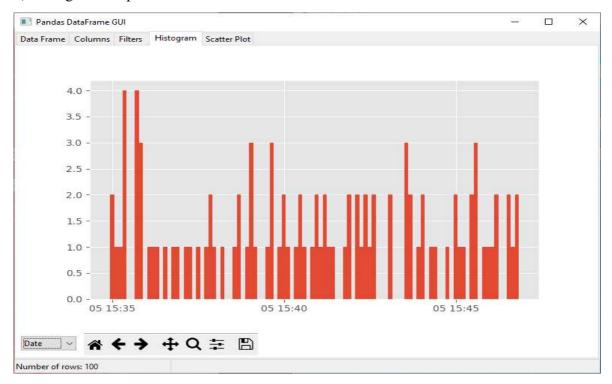
3) Enter Hashtag Screen:



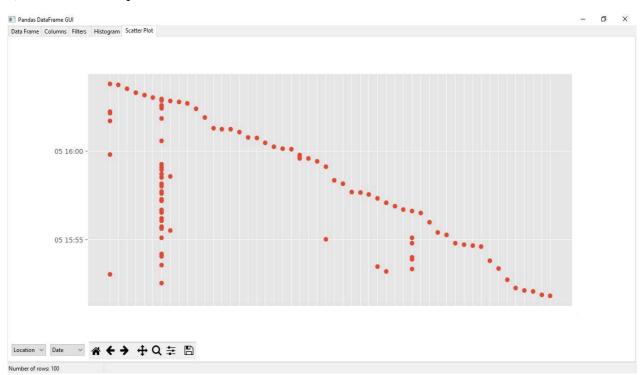
4) User, Tweet, Location, Date Output:



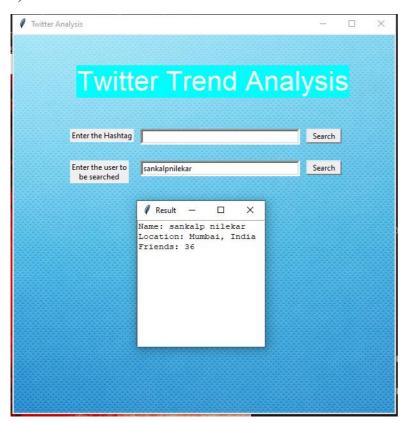
5) Histogram Output:



6) Scatter Plot Output:



7) User Search Screen:



CONCLUSIONS

9.1 Contributions

The main contribution of this project report is to suggest a new unique way to analyze the trends in online social networks. We have identified some of the features, which can help develop a model, which can be used to classify "trends" and "non-trends" in very early stages. We have also developed a highly scalable and efficient model to filter noise from tweets. Also, this prediction model is generic enough to be applied to any social media network, which has a connection among users. Thus, we can conclude that by constructing evolving graphs for different topics and observing several topological properties of the graph we can distinguish "trend" from "non-trends" topics.

9.2 Future Work

We have certainly established the benefits of constructing a social graph of twitter. So resolving more relations for the users in the graph can be useful and can improve the performance of the model. The model's success depends on the topic of wise clustering of tweets. Currently, we have used simple clustering which not very "strict", other better clustering algorithms can be used. Storing and processing graphs have been the real challenge and bottleneck of the whole pipeline. It's necessary to improve this step by exploring better ways to do the same. Other data structures and algorithms can be explored to process the graph faster.

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