

# A Sequential Prediction Model for Topic Popularity in Social Media Networks

Vimal Kumar Gangwar<sup>1</sup>, Dr. Tarun Kumar<sup>2</sup>

<sup>1</sup>Student, Computer Science and Engineering, Radha Govind Group of Institutions Meerut, India

<sup>2</sup>Head of Department, Computer Science and Engineering, Radha Govind Group of Institutions Meerut, India

\*\*\*

**Abstract** - It is extremely popular to identify trendy topics, which can profit numerous assignments including topic prediction, the direction of popular sentiments, for future precaution in upcoming problems, etc. However, at point, peoples might need to know when to make a topic popular. In this paper, we address this issue by presenting a Sequential Prediction Model for User Topic Selection (UTS) which models clients' practices of posting messages and comments. The UTS model considers clients' interests, companion circles, and startling occasions in online interpersonal organizations. Likewise, it considers the continual fleeting displaying of points, since themes are changing constantly after some time. Besides, a weighting plan is proposed to alterations in the changes of topic popularity. At long last, the trial prediction comes about on true informational collections and generates the prediction graph based on the user's thoughts and interests which will reveal the success of our proposed models, and point topic popularity prediction.

**Keywords**— User topic Selection (UTS), Rehotting topic, online social network, Graphical design, EM algorithm

## 1. INTRODUCTION

Online web-based life, for example, Twitter and Weibo give major stages for the person to pass on their opinion, exchange their perspectives, and offer their experiences. On these stages, individuals revive their status concerning various subjects, crossing from "Twitter", to "Facebook", and this is given quickly to their buddies. This essentially strengthens singular exchange and interest. Other than empowering interchanges among clients which contain rich information like related to banks, schools, and government affiliations, etc. Various affiliations are enthused about continually mining and separating this client produced social data as a result of the accompanying reasons. To begin with, the social information contains the interests, concerns, and assessments of their clients, and offers pointers to relationship to upgrade their items or organizations. Second, social data irrefutably contains significant market bits of information for the affiliations. The fundamental foundation of these significant level applications relies upon subject watching and comprehension. Specifically, affiliations may need to: track the improvement of any perceived relevant subjects about them; and be instructed of any new rising focuses which are fast gathering power in microblogs. That has found uncommon achievement in building subject

models of static substance. Variations of PLSA and LDA have been proposed for the web and dynamic theme demonstrating. An alternate line of related exploration is on word-reference learning and non-probabilistic system.

## 2. LITERATURE SURVEY

With the coming of advanced databases and correspondence systems, immense storehouses of literary information have opened up to a huge open. Today, it is one of the incredible difficulties in the data sciences to create clever interfaces for human {machine cooperation which bolster PC clients as they continue looking for applicable data. Even though the utilization of expounding ergonomic components like PC illustrations and representation has demonstrated to be amazingly productive to encourage and upgrade data get to, progress on the more essential inquiry of machine insight is at last important to guarantee significant advancement on this issue. With the goal for PCs to communicate all the more normally with people, one needs to manage the expected inner conflict, uncertainty, or even dubiousness of client demands and needs to perceive the distinction between what a client may state or do and what she or he really implied or planned. One regular situation of human {machine connection in data recovery is by characteristic language questions: the client details a solicitation, e.g., by giving various catchphrases or some freestyle text, and anticipates that the framework should restore the important information in some amiable portrayal, e.g., in the type of a positioned rundown of pertinent reports. Numerous recovery techniques depend on basic word coordinating methodologies to decide the position of the importance of an archive concerning a question.

However, it is notable that the exacting term coordinating has serious downsides, fundamentally because of the indecision of words and their unavoidable absence of exactness just as because of individual style and individual contrasts in word utilization. Latent-Semantic-Analysis (LSA) is a way to deal with programmed ordering and data recovery that endeavors to defeat these issues by planning reports just as terms to a replica in the so-called latent semantic environment. LSA as a rule takes the (high dimensional) vector space portrayal of reports dependent on term frequencies as a beginning stage and applies a measurement diminishing direct projection. The particular type of this planning is dictated by a given record assortment and depends on a Singular Value Decomposition (SVD) of the

relating term/report lattice. The overall case is that similarities between reports or among archives and inquiries can be all the more dependably assessed in the diminished dormant space portrayal than in the first portrayal. The method of reasoning is that reports which share often co-happening terms will have a comparative portrayal in the inactive space, regardless of whether they share no terms practically speaking.

LSA along these lines plays out a type of clamour decrease and has the expected benefit to identify equivalent words just like words that allude to a similar theme. In numerous applications, this has demonstrated to bring about increasingly hearty word preparing. In spite of the fact that LSA has been applied with wonderful accomplishments in various spaces including programmed ordering (Latent Semantic Indexing, L.S.I), it has various deities, for the most part, because of its inadmissible measurable establishment. The essential objective of this paper is to introduce a novel way to deal with LSA and factor investigation {called Probabilistic Latent Semantic Analysis (PLSA) {that has a strong factual establishment since it depends on the probability standard and uses an appropriate generative model of the information. This infers specifically that standard strategies from insights can be applied for questions like adjusting, model blend, and multifaceted nature control. Likewise, the factor portrayal got by PLSA permits to manage polysemous words and to unequivocally recognize different implications and different kinds of word use.

### 3. PROBLEM DEFINITIONS

This paper tends to the issue of topic trendy expectations. As appeared in Fig-1, we might think through the accompanying 2 methods to manage the topic re-hotting expectation issue.

- The separate demonstrating method segments the full time-space into adjacent non-covering time display, and after that utilizes the readied information (portrayed blue dotted line) for envision might be the subject will popular at whatever point windows (i.e, in session of  $t_5$  to  $t_6$ ). Despite the way that this methodology is viably reasonable, it can't foresee the specific time centers around re-hotting a given point. Also, it is hard to depict the changing examples of topics in a fine-grained way.
- Constant demonstrating strategies describes the contents will be altering continuously in particular time bound. Considering an accomplished information (outlined Red continuous line), at specific period centers this will be predicted. When the theme will re-hot, e.g., at the time point  $t_5$ . This framework will able to prediction as perfectly at given time bound. (Outlined red broken line).

### 4. LIMITATIONS OF PROJECT

There are numerous gigantic preliminaries to mention this issue. Initially, it's nontrivial to validate the matter of point re-hotting assumption and genuinely techniques for content selection. Also, this will be frightfully hard to precisely get a lucky time point to re-hotting a provided subject. Finally, the content rehotting selection techniques are difficult to suggest

### 5. EXISTING SYSTEM

As online long range interpersonal communication develops, there has been expanded enthusiasm to use the fundamental system structure just as the accessible data on social media to improve the data needs of a client. In this paper, we center on improving the analysis of data assortment from the area of a client in a unique informal organization. We acquaint testing based calculations with effectively investigating a client's interpersonal organization regarding its structure and to rapidly rough amounts of premium. We present and break down variations of the essential inspecting plan investigating connections over our examples. Models of brought together and dispersed informal communities are thought of. We show that our calculations can be used to rank things in the area of a client, expecting that data for.

### 6. PROPOSED SYSTEM

The discrete demonstrating strategies separate the full area into touching non-covering time windows, at that point utilize the prepared data (portrayed Blue dotted line) predict subject can be re\_hot inside ( all through sum  $T_5$  to  $T_6$ ). Also, it's challenging in clarification of dynamic outlines in subjects extremely smooth manner.

The consistent demonstrating approach shows that topics are habitually changing in the time area. In view of the prepared data (delineated in Red continues line), it assume specific period to focuses when the theme will re\_hot, such as, period points  $t_5$ .

Generally the used methodology may anticipate in re-hotting period focuses on an all-encompassing measure of time (delineated as red-spotted lines) as opposed to just a future time window.

### 7. FUTURE USAGE

This will contains the rich information's towards the organizations such as banks, colleges, government organizations etc. The mining of useful data which will reflect the user interest and opinions and also help the organizations to improvement in their products and services.

- Used in digital marketing through research in users opinions.
- Disaster management systems

- Medical field sections
- Source of income through blogging
- Educations Sector
- Update current affairs\Media Sector

**8. DESIGN STRUCTURE**

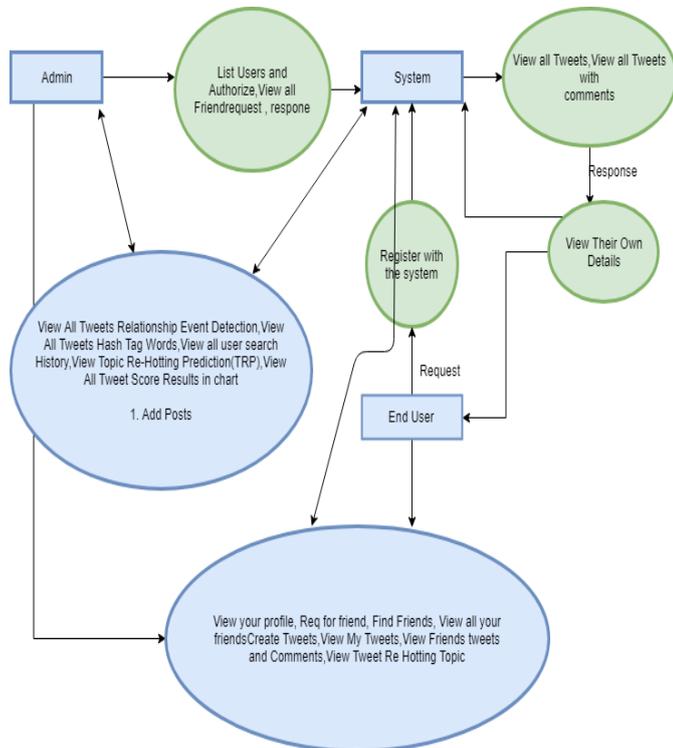
**EM Algorithm:**

The Expectation-Maximization (EM) algorithm is a comprehensively utilized strategy to figure the most extreme probability gauges, which benefits an assortment of inadequate information issues [19]. The EM calculation is initially proposed by Dempster, Laird, and Rubin [20]. For models with potential factors, it is hard to locate the greatest probability specifically. The EM calculation answers such issues. As an iterative calculation, there are two stages in every emphasis of the EM calculation — the Expectation (E) step and the Maximization (M) step.

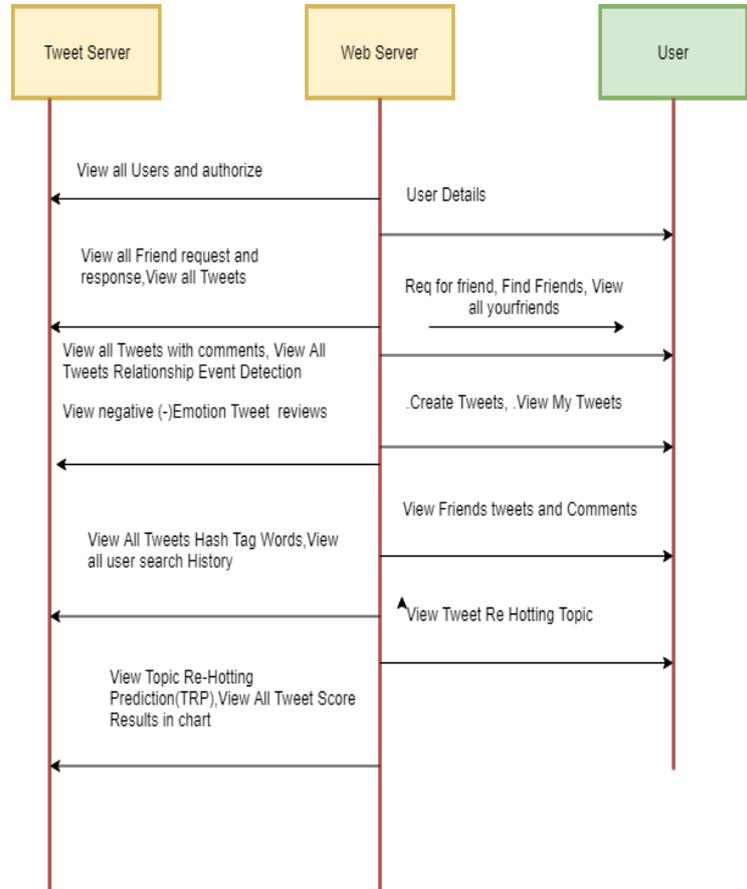
**The Event-driven UTP Model:**

The Event-driven UTP (E-UTP) show gives careful consideration to the impact of an unforeseen event on topics. A surprising event is an external event, for example, a fear monger assault, an illness flare-up, or an auto collision.

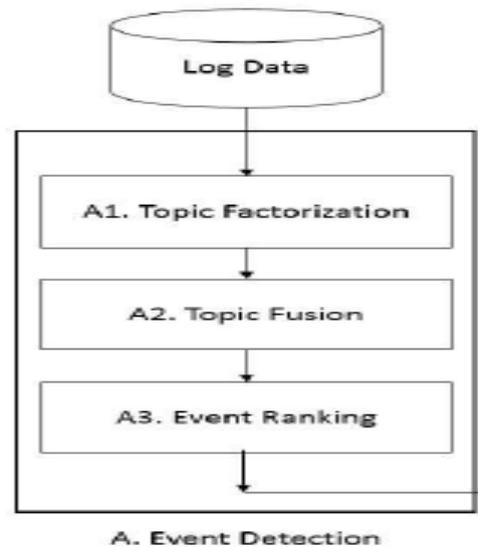
**Data Flow Structure**



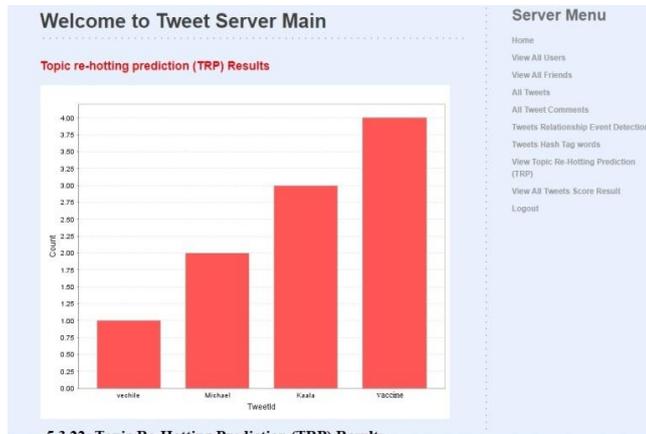
**Sequenced Diagram**



**Implementation Method**



• **Output Graph**



**9. CONCLUSIONS**

As prediction of trendy hot topic in accurate process is not much easier, this will be solve by using the 3 main components of the user activities such as their friends circle, different concepts of content discussions, & unforeseen activities carried out by users. The UTS techniques will examine all these components one by one to get prediction. And also E-M algo is used to provide help in accuracy of prediction strategy in the trendy topics.

Here we show the execution of the UTS user topic selection with the real datasets by generating the real time graph, this will help to analyse the popularity of the hot topics. For prediction of trendy topic will be more perfectly in upcoming period and this proposing techniques will decrease the issues related to information of online network.

**10. REFERENCES**

- H. Kwak, C. Lee, H. Park, and S. Moon, "What is Twitter, a Social Network or a News Media?" in Proceedings of the 19th International Conference on World Wide Web. New York, NY, USA: ACM, 2010, pp. 591–600, 2010  
<https://pdfs.semanticscholar.org/225b/1d2a863fe488f488f87d4a91c3038b3d02b8d894.pdf>
- P. Megha Nagawad 1, Prof. Vandana V. Navale2 "Survey on a temporal model for topic re-hotting prediction in online social network" in the journal of open access international journal of science & engineering, vol. 3, pp. 8-10, 2018
- [http://www.oaijse.com/VolumeArticles/FullTextPDF/314\\_3.SURVEY\\_ON\\_A\\_TEMPORAL\\_MODEL\\_FOR\\_TOPIC\\_RE-HOTTING.pdf](http://www.oaijse.com/VolumeArticles/FullTextPDF/314_3.SURVEY_ON_A_TEMPORAL_MODEL_FOR_TOPIC_RE-HOTTING.pdf)
- B. Quost & T. Denoeux, "Clustering & Classification of Fuzzy Data Using the Fuzzy EM Algorithm", Journal of

the Royal Statistical Society, vol. 286, no. 1, pp. 134-156, 2016

- X. Zhang, X. Chen, Y. Chen, S. Wang, Z. Li, and J. Xia, "Event Detection and Popularity Prediction in Microblogging", Neurocomputing, vol. 149, pp. 1469-1480, 2015
- P. Velardi, G. Stilo, A. E. Tozzi, and F. Gesualdo, "Twitter Mining for Fine-grained Syndromic Surveillance," Artificial Intelligence in Medicine, vol. 61, pp. 153-163, 2014
- <https://pdfs.semanticscholar.org/6f2b/555fffeabd65442c47b32daf191487d12355.pdf>
- K. Liu, J. Xu, L. Zhang, Z. Ding, and M. Li, "Discovering Hot Topics from Geo-tagged Video," Neurocomputing, vol. 105, pp. 90-99, 2013  
<http://or.nsf.gov.cn/bitstream/00001903-5/90947/1/1000003760183.pdf>
- W. Li and A. McCallum, "Pachinko Allocation: DAG-structured Mixture Models of Topic Correlations," in Proceedings of International Conference on Machine Learning. ACM, pp. 577-584, 2006