

# A Novel Multi Context Prediction and Mining Model on Online Opinions Evolutions against Food Hazard

<sup>1</sup>N.Hamsaleka, <sup>2</sup>Dr.V.Kathiresan

<sup>1</sup>Research Scholar, Department of Computer Science, Bharathiar University, Coimbatore, Tamil Nadu, India

<sup>2</sup>Head of the Department, Department of Computer Applications (PG), Dr. SNS Rajalakshmi college of Arts and Science, Coimbatore, Tamil Nadu, India

## ABSTRACT

Proliferation of location based social Network yields the variety of opinion related to food quality and hazards as data. Many data analysis technique has been proposed to determine specific fact. Despite of various factors, Analysis of collective inference of food related review on multi context stands primary important of the current research. Though we propose a novel technique entitled as Multi Context Prediction and Mining Model on online opinions Evolutions. It exploits the correlated multi context based on the conceptual similarity at different time period in terms of time series analysis and feature representation. Further it reduces the invariance of feature evolution and concept evolution to more extent. This model involves complex relationship determination among the instances. Variation of related instance evolved over time can be easily extracted. It works as multi context evolution inference model. Additionally opinion adaptation model is been defined to categorize the Target opinion on particular event using probability distribution model or markov random field. Also it helps in reducing the labelling efforts of the opinion by leveraging labelled data from various criterias. The extensive experimental results prove that proposed model outperforms the state of art approaches in terms of precision, Recall and F Measure.

**Keyword:** Food Hazard Prediction, Opinion Evolution, Opinion Inference, Online Review, Opinion Mining

## I. INTRODUCTION

Nowadays ample opinions related to food are been posted in social media and location based service applications [1]. Mining and prediction of the Target opinion is critical research challenges. For instance Facebook, Yelp and Foursquare are inundated by the opinion overloading on the various aspects of their interest. Opinions are inferred independently to food quality estimation and hazard estimation [2]. In existing many supervised and unsupervised model has been derived to predict the different label to large stream of the opinions based on the correlation and covariance. Besides component analysis, opinion involves multiple evolution attributes representing

different food volumes. In this paper, we analysis food related review through collective inferences on multi context. It exploits the correlated and covariance multi context based on the conceptual similarity and semantic similarity at different time period as time series analysis [3]. The opinion is represented as features. The proposed analysis involves complex relationship determination in order to reduce the invariance of feature evolution and concept evolution among the instances.

Variation of related instance evolved over time can be easily extracted using vector space model which works as multi context evolution inference model. Additionally opinion adaptation model is been

defined to categorize the opinion on particular target event on the various distributed opinions [4]. It easily adapts the opinion of large variety into specified target category using probability distribution or using Markov Random Field [5]. The remainder of the paper is organized as follows: Section 2 discusses the related works in opinion evolution exploiting under multiple preferences; Section 3 discusses the proposed technique in terms of multi context prediction and mining technique on the evolving opinion. Section 4 presents the experimental results. Finally Section 5 discusses conclusions.

## II. RELATED WORK

There exist many techniques to mine the food hazards from the online reviews through various opinion mining techniques. Each of those techniques follows some kind of processing to predict the target opinion which is described as follows

### Aspect Preference based opinion Classification

It is considered as aspect based opinion classification model, model utilizes the Joint Aspect/Sentiment (JAS) model in order to jointly extract aspects and aspect-dependent opinion features from food reviews. An aspect-dependent feature refers to the aspect-specific opinion words along with their aspect-aware context polarities with respect to a specific aspect of the opinion. The extracted aspect-dependent features is been applied to a series of aspect-level opinion mining tasks, including implicit aspect identification, aspect-based extractive opinion summarization, and aspect-level classification [6].

## III. PROPOSED MODEL

In this section, we describe the Multi Context Prediction and Mining Model on online opinions Evolutions against Food Hazard using Opinion Classification is as follows

### Feature Representation

The Feature representation is a process used to detect the features in the online review from the dataset after preprocessing of the information such as elimination of stop words and stemming process. The feature is presented using frequency based extraction technique which yield feature of the opinion in terms of specifying the threshold as any number to determine height number terms to be treated as feature.

Let us represent Feature Vector  $F = \{f_1, f_2, f_3...f_n\}$

### Evolution Prediction

The Evolution is determined for the feature of upcoming opinion that different instances may be at different stages of their time cycles at time which it is computed for concept extraction to the term of the instance or semantic extraction to the instance. The extracted concept and semantic of the word or term determines same to already detected feature, it is merged else it will be represented as new feature

Let  $f_7 = \{c_1, c_2, c_3\}$

$c_1, c_2, c_3$  implies conceptual meaning for  $f_7$

If ( $c_1 = f_2$ )

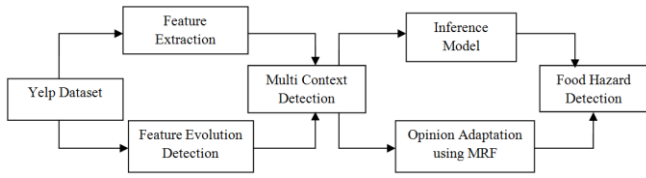
Merge  $f_7$  into  $f_2$

Else

Represent  $f_7$  as new feature in the feature vector

### Inference model based multi context

Conventional inference approaches usually require dependency between different instances based on the multi context of the opinions. The inference for each instance is performed independently on the food concepts. Collective inference approaches assume that the instances are related with directly linked instances in the opinion. Different instances have different influence on the target instance, which cannot be reflected from the directly linked relationships. The figure 1 represents the architecture of the proposed model.



**Figure 1:** Architecture of the Proposed Multi context Prediction and Opinion Mining model

**Opinion Adaptation Model using Markov Random Field**

The opinion Adaptation model iteratively predicts the target value. These methods rely on the assumption that deep learning can successfully learn the desired transferable representations for opinion adaptation using markov Random Field. It is applied on joint probability condition on pairwise property estimation. It determines the opinion dependency on the specified condition or some hidden condition in order to determine target opinion.

**IV. EXPERIMENTAL RESULTS**

In section, we describe the experimental results of the proposed framework against the existing approaches on the yelp dataset. Yelp dataset mostly concern same set of entities concerning feature evolution. The data are of realistic sizes for distinct instance of the particular coverage problem. To build effective multi context opinion prediction model, we have generated feature and merged the feature which is evolved in some aspects. Target opinion adaptation is also included using pair wise similarity measures on MRF model. To validate the effectiveness of the proposed model, we use different measure such as precision, Recall, F Measure. Precision is the fraction of true matching pairs within the set of classified matching pairs.

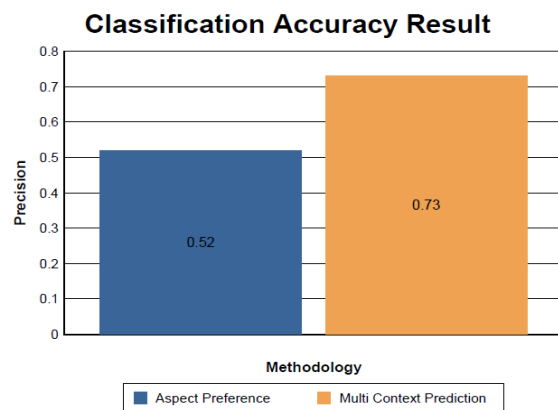
In Figure 2, the performance of the proposed model evaluated in terms of precision. To get an intuitive sense of the matching quality, we show some examples of matching pairs on the different context.

On the other hand, the proposed algorithm is much more efficient, but it produces an approximate solution. Recall is the fraction of true matching pairs found by the classifier within the set of all true matching pairs. The Figure 3 describes the performance outcome of the proposed model in terms of recall value.

Multi context opinion prediction is some cases leads to class posterior condition due to the fact that every selected opinion has to meet the efficiency threshold, reducing the set of available candidate opinion. This verifies that the proposed models can represent high transferable feature to

Enable opinion adaptation.

F measure is used to detect the test the accuracy of the precision and recall through computation of weighted harmonic mean. The Figure 4 represents the performance outcome of the proposed model. The proposed model is flexible and effective to apply to the food hazard and food quality detection task even based on opinion trajectories. Table 1 describe the effectiveness of the proposed model.



**Figure 2:** Performance Evaluation of different mechanism in terms of Precision

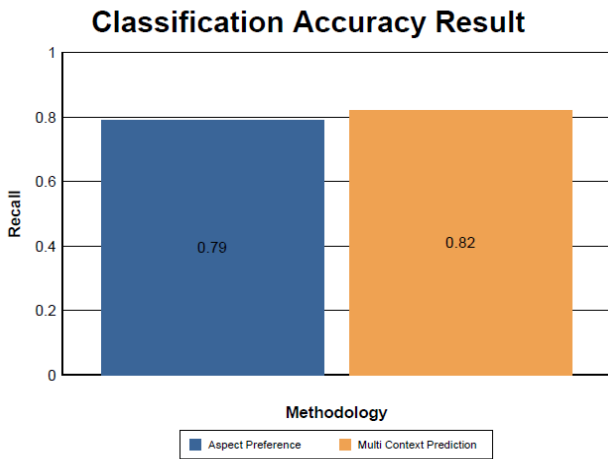


Figure 3: Performance Evaluation of different mechanism in terms of Recall

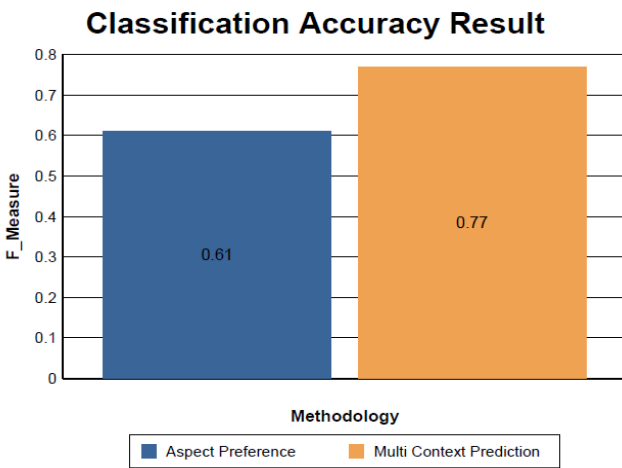


Figure 4: Performance Evaluation of different mechanism in terms of F measure

Table 1: Performance Evaluation on proposed model on different measures

Technique	Precision	Recall	F measure
Aspect Preference based opinion Classification – Existing System	0.52	0.79	0.61
Multi Context Prediction and Mining Model on online opinions Evolutions – Proposed System	0.73	0.82	0.77

All the contextual information on the multi context is represented in the same target opinion prediction in order to make it convenient to analyze the association among different contexts of various sizes of opinions.

### V. CONCLUSION

We designed and implemented a Multi Context Prediction and Mining Model on online opinions Evolutions through strong analysis of collective inference of food opinion on multiple opinion pools. The proposed model has exploited the various food opinion of the user by involving complex relationship determination among instance which is correlated and uncorrelated multi context. The Content analysis is based on conceptual similarity at different time period gives the more prediction accuracy. The Exploitation is carried out as time series analysis through feature representation models. Implementation Results have proved that invariance of feature evolution and concept evolution has been reduced. Finally it has been proved that categorization of opinion to the target is achieved effectively.

### VI. REFERENCES

- [1] Y. Liu, X. Huang, A. An, and X. Yu, “Modeling and predicting the helpfulness of online reviews,” in Proc. 8th Int. Conf. Data Mining, 2008, pp. 443–452.
- [2] P. Tsaparas, A. Ntoulas, and E. Terzi, “Selecting a comprehensive set of reviews,” in Proc. 17th ACM SIGKDD Int. Conf. Knowl. Discov. Data Mining, 2011, pp. 168–176.
- [3] Bo Liu, Yanshan Xiao, Philip S. Yu, Zhifeng Hao, Longbing Cao "An Efficient Approach for Outlier Detection with Imperfect Data Labels" IEEE Transactions on Knowledge and Data Engineering in Volume: 26, Issue: 7, July 2014
- [4] R. Jin and G. Agrawal. Efficient decision tree construction on streaming data. In Proceedings

of the 9th ACM SIGKDD International Conference on Knowledge Discovery and Data Mining, pages 571–576, 2003.

- [5] M. Masud, J. Gao, L. Khan, J. Han, and B. Thuraisingham. Classification and novel class detection in concept-drifting data streams under time constraints. *IEEE Trans. Knowledge and Data Engineering*, 23(6):859–874, 2011.
- [6] K. Zhang, V. W. Zheng, Q. Wang, J. T. Kwok, Q. Yang, and I. Marsic, “Covariate shift in hilbert space: A solution via surrogate kernels,” in *Int. Conf. Mach. Learn.*, 2013.
- [7] Dingqi Yang, Daqing Zhang, and Bingqing Qu. Participatory cultural mapping based on collective behavior data in locationbased social networks. *ACM Trans. Intell. Syst. Technol.*, 7(3):30:1– 30:23, 2016.
- [8] J. Tang, S. Wu, J. Sun, and H. Su, “Cross-domain collaboration recommendation,” in *ACM SIGKDD Int. Conf. Knowl. Discovery Data Mining*, 2012.
- [9] W. Yu, R. Zhang, X. He, and C. Sha, “Selecting a diversified set of reviews,” in *Proc. 15th Asia-Pacific Web Conf.*, 2013, pp. 721–733.
- [10] L. Yu, P. Cui, F. Wang, C. Song, and S. Yang. From Micro to Macro: Uncovering and predicting information cascading process with behavioral dynamics. In *ICDM*, 2015.