

# **A Methodology for Cyber Crime Identification using Email Corpus based on Gaussian Mixture Model**

V.Sreenivasulu

Department of Computer Science and Engineering  
Gandhiji Institute of Science and Technology,  
Krishna District - Andhra Pradesh - India

R. Satya Prasad, PhD

Department of Computer Science and Engineering  
Acharya Nagarjuna University - Guntur,  
Andhra Pradesh - India

## **ABSTRACT**

The area of crime investigation has extended its roots to cyber media and has emerged exponentially with the technological strides. Among the various media used in Digital Forensics, Email Forensics took up the leading segment. In order to investigate the cyber crimes, there is an immense need to analyze the bulky email gatherings forensically. Data mining methods help in analyzing these large collections of data. Mixtures of data mining models along with the related methodologies are proposed in this paper to facilitate the email forensic assessor. The Performance is evaluated using False Rejection Ratio (FRR) and False Acceptance Ratio (FAR).

## **Keywords**

Email Forensics, Data Mining, Digital Forensics, Word Net, Query Analysis.

## **1. INTRODUCTION**

The rapid expansion of science and technologies has helped the mankind in witnessing new methods for leading a better quality of life. At the same time these technological updates have paved new approaches for the criminals and anti-social elements in indulging the criminal activities using the cyber media as a medium. To prevent these activities, effective methods are needed. Among the various methods of identifying the cyber criminals, analysis of the emails is considered to be more effective. The practice of emails for the fraudulent actions is also emerging with an elevated speed. Forensic Analysis of these emails can probe or prove a crime committed. Email forensics can appropriately be defined as the usage of focused techniques used for the collection, protection and analysis of emails with a vision to submit the evidences in a court of law. As the usage of internet technologies are expanding, the usage of emails has been increased and investigating these bulk emails is a great task for the forensic examiners. Therefore new methods are to be evolved for the effective analysis of these emails to uproot the inherent information concealed in the emails.

The core idea behind the digital forensics procedures is to identify the relevant information from the digital media and which can be used as evidence in proving a crime. To recognize and interpret the meaningful information hidden in the emails, Information Retrieval (IR) plays a vital role. Information Retrieval systems recognize the relevant information by retrieving the data based on the user's queries. In this paper we recommend a system to retrieve the most related emails from the bulk email collections of emails and exhibit these emails to the forensic examiners for better understanding. Every piece of information is of core essence in the investigation process, so it

is needed to convert the information retrieved into a meaningful format so that it will be the utmost usage for the Forensic examiners. Every forensic information retrieval systems work with the aim towards high recall ratio. The present system addresses the issue of high recall using the concepts of interpreting the semantic meaning of the words within the emails for which in the present paper the concepts of ontology i.e. Word Net is used.

In order to have an effective examination, the forensic examiners should identify the emails which have a high degree of matching with the relevant information; this is assumed to be the primary goal of any system which aims at High recall. In this paper, we present a novel methodology, where the emails from a particular user under examination are scanned and the text from each of these emails is extracted and the contents are indexed. This indexing list is given as input to the semantic analyzer using Word Net and the related synonyms of the words are listed. This list is maintained in a book called code book. Further, for each of the words, the grammar is also extracted; the relevant acronyms are expanded using the word Net. Concepts of clustering summarization are used. The rest of the article is organized as follows, in Section-2; the related work in this area is highlighted, in Section-3; the dataset considered is presented along with a sample from the corpus. The methodology is highlighted in Section-4 of the paper; and Section 5; concludes the paper.

## **2. RELATED WORK**

The restriction of the message sharing and content sharing is a concern because of the deceptive activities taking place all around the globe. Therefore, it is needed to counter attack these fraudulent activities to curtail the criminal activities. Lot of research has been reported in the literature regarding methodologies proposed to curb the fake emails and disrupting the activities. Data mining techniques of classification and clustering have given unbelievable results in email forensic analysis as explained in [1][2][3]. Among the available filters, Naïve Bayesian is a basic Standard Statistical approach on probability proposed by Sahami et al [4]. The algorithm assumes the classification of new Email by identifying it as spam or legitimate. It achieves the features using a training set that has already been pre classified correctly and then checks for specific work that appears in the email. The indication of high probability says that the new one is a spam email. In recent years the authorship identification has been used in diverse number of application areas some examples of identifying authors in written literature, in the program source code [5] and in forensic analysis for criminal cases. Some techniques involve identification of Spoofed E-Mails [6] based on various techniques of SPF (Sender Policy Framework), Sender ID, Domain Keys Identified Mail (DKIM). In most of the



## 4.1 Algorithm

The proposed architecture of the model is presented as follows.

### Step 1:

The email datasets are extracted from the ENRON dataset and preprocessed to eliminate the missing data and eliminate the stop words. The preprocessed data is given as an input to the semantic tool, WORDNET. The purpose of this tool is to analyze the grammar, identify the term index, acronyms and this data is also used for identifying the different synonyms for a given keyword. The synonyms in this paper are considered only for the adjectives. A code book is generated for the group of adjectives and the indexing is carried out for each word in the code book. This indexed data for email is given as input to the Gaussian Mixture Model (gmm). The main advantage of considering gmm is that the patterns inside the data can be clearly estimated. It is symmetric model and can be very useful for analyzing huge data sets. This mixture model is very effective for modeling the data having very low index and very high index. The probability function of the Gaussian Mixture Model is given by

$$N(X|\mu, \Sigma) = \frac{1}{(2\pi)^{D/2}} \frac{1}{|\Sigma|^{1/2}} \exp\left\{-\frac{1}{2}(x - \mu)^T \Sigma^{-1}(x - \mu)\right\} \dots (1)$$

### Step 2:

The indexed data is given as input to the gmm presented in equation (1) and the corresponding probability density function (pdf) is estimated. The process is repeated for all the data in the email and all the emails in the dataset. This is called training data. For testing purpose the same procedure is adopted and pdf of the query email is generated. The pdf is compared with the pdfs in the training data using KL divergence (Khaul Loullie Formulae). It is used to identify the relation between the email in the database and the email which happens to be false email i.e., the suspected email. MathCAD is used for the calculation purpose where the log of A and B are first calculated and then the integration is carried out. The MathCAD is used to perform the mathematical calculation of integration, differentiation and volume integrals. The alternative to K L Divergence is identifying the likelihood estimate. The formulae for calculating the KL divergence is given by

$$KL(A, B) = \int A(x) \log\left(\frac{A(x)}{B(x)}\right) dx \dots (2)$$

Where A(x) is the corpus of a person x and B(x) is the email which is under consideration.

### Step 3:

For each query email, the relevant PDF is retrieved. The process is repeated for all the query emails to find the best fit. The results derived are presented below.

### Step 4:

To evaluate the performance of the given model, we used metrics like False Acceptance Rate (FAR) and false Rejection Rate (FRR).

**Table 1 The accuracy, FAR and FRR of Emails**

Trait	Algorithm	Acceptance Rate (%)	FAR (%)	FRR (%)
Email-set1	GMM	88.70	0.07	0.93
Email-Set2	GMM	91.05	0.064	0.936

The above outputs evaluated are compared using Acceptance Rate and False Acceptance Rate. The FAR is calculated using

$$\frac{[(Dataset\ under\ consideration) - (Total\ Number\ of\ related\ words)]}{Size\ of\ total\ data\ set} \times 100$$

The acceptance rate is calculated as

$$\left[ \frac{Total\ Number\ of\ items}{Number\ of\ items} \times 100 \right]$$

## 5. CONCLUSION

In this paper a novel methodology for email forensics is highlighted using the concepts of data mining, semantic ontologies and Gaussian mixture model. The outputs derived are tested against accuracy using metrics like FAR and FRR. The results derived are presented above and revealed that the proposed methodology possesses high false acceptance rate and low false rejection rates. This methodology can be very useful in identifying from email corpus and thereby helping to identify the law breaker.

## 6. REFERENCES

- [1] Rachid Hadjidj *et al* "Towards An Integrated Email Forensic Analysis Framework", Digital Investigation 5, pp.124-137, 2009.
- [2] S.Appavu alias Balamurugam, Dr.R.Rajaram,"Data mining techniques for suspicious email detection: A comparative study", IADIS European Conference Data Mining, 2007.
- [3] D.v. Chandra Sekhar and S. Sagar Imambi," Classifying and Identifying of Threats in Email Using Data Mining Techniques", Proceedings of the International MlitiConference of Engineers and Computer Scientists Vol.1,I IMECS,19-21 March 2008,Hong Kong.
- [4] Sahami *et al* "A Bayesian Approach to Filtering Junk Email In Learning for Text Categorization" – papers from the AAAI Workshop, pp. 55-62, Madison Wisconsin AAAI Technical Report WS-98-05, 1998.
- [5] Gray *et al*," Software forensics: Extending authorship analysis techniques to computer programs", Third biannual conference of the international Association of Forensic Linguists (IAFL'97),1997.
- [6] Dhanalakshmi R,L.Kavisankar,C.Chellappan"Enhanced Email Authentication Against spoofing Attacks To Mitigate Phishing, European Journal of Scientific research Vol . Issue 2011
- [7] Marwan Al-Zarouni. (2004). "Tracing E-mail Headers", Proceedings of Australian Computer, Network & Information Forensics Conference on 25th November, School of Computer and Information Science, Edith Cowan University Western Australia 2004, pp. 16-30.

- [8] eMailTrackerPro, <http://www.emailtrackerpro.com/>
- [9] EmailTracer, <http://www.cyberforensics.in>
- [10] Adcomplain, <http://www.rdrop.com/users/billmc/adcomplain.html>
- [11] Aid4Mail Forensic, <http://www.aid4mail.com/>
- [12] AbusePipe, <http://www.datamystic.com/abusepipe.html>
- [13] AccessData's FTK, <http://www.accessdata.com/>
- [14] EnCase Forensic, <http://www.guidancesoftware.com>
- [15] FINALeMAIL, <http://finaldata2.com>
- [16] Sawmill-GroupWise, <http://www.sawmill.net>