Chap4#5#2: Machine Learning for Anomaly-based Spam Detection

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Topics to study in Chapter 4

 Machine learning for Anomaly Detection: Definition of an anomaly. Types of Anomalies or outliers in machine learning. Motivation for machine learning for anomaly detection.

Data Visualization. Supervised, Unsupervised and Semi-supervised Learning methods for Anomaly Detection.

Applications of Anomaly Detection: Intrusion detection, Fraud detection, Health monitoring, Defect detection, and lastly Spam detection. Intrusion Detection with Heuristics.Goodness-of-fit. Host Intrusion Detection.

Network Intrusion Detection. Web Application Intrusion Detection.

Overview of Machine learning Approaches for Anomaly Detection:

Distance-based, Clustering-based and Model-based Approaches. Algorithms for Distance and Density-based approaches, Rank-based approaches, Ensemble Methods Algorithms for Time Series Data. Deep Learning for Anomaly Detection. Behavioural-based Anomaly Detection [8 hours]

Topics in Handouts#1, #2, #3

- 1. Email Basics, Spam Detection Basics, and ML-based Spam Detection Basics
- 2. Classical ML-based Spam Filtering
- 3. Deep Learning-based Spam Filtering

ML-based Methods used for Email Spam filtering

Clustering-based ML Spam filtering. Clustering....

• is a type of approach used in dividing objects or case examinations into comparatively similar collections known as clusters.

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- Density-based clustering and K-nearest neighbours (kNN).



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- Two types of clustering methods that have been used for Spam classification as such.
- Density-based clustering and K-nearest neighbours (kNN).
 - density based clustering implemented in² showed the capacity to process encrypted messages too, thereby upholding privacy confidentiality.



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 - this property is very vital as in the actual scenario, nearly all of the applied data disobey the standard hypothetical postulations made (such as Gaussian mixture, linearly separable, and others).

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- is termed as a lazy learner since the training data points is not used by it to perform generalization. That is, there is no obvious training stage and if it exists it is extremely small.

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- the implication is that the algorithm has a moderately speedy training phase.
- however, the entire training data is required throughout the testing phase as decisions are made based on the complete training data set.

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ML-based Spam filtering: Clustering: kNN algorithm pseudocode

Here Neighbours(d) return the k nearest neighbours of d, Closest (d, t) return the closest elements of t in d, and testClass(S) return the class label of S.

Algorithm 1 kNN Algorithm for Spam Email Classification 1: Find Email Message class labels. 2: Input k, the number of nearest neighbors 3: Input D. the set of test Email Message:

- 4: **Input** *T*, the set of training Email Message.
- 5: *L*, the label set of test Email Message.
 6: Read DataFile (TrainingData)
- 7: Read DataFile (TestingData)
- 8: **for** each d in D and each t in T **do**
- 9: Neighbors(d) = {}
- 10: if |Neighbors (d) | < k then
- 11: Neighbors(d) = Closest (d, t) \cup Neighbors(d)
- 12: end if
- 13: if $|\text{Neighbors}(d)| \ge k$ then
- 14: $restrain(M, x_j, y_j)$
- 15: end if
- 16: end for 17: return Final Email Message Classification (Spam/Valid email)
- 18: end

Figure: KNN based Clustering for Spam detection

¹S. Zhu, W. Dong, W. Liu, Hierarchical reinforcement learning based on KNN classification algorithms, Int. J. Hosp. Inf. Technol. 8 (8) (2015) 175–184.