











## 8. Conclusion

In this paper we compared the three main intrusion detection algorithms: FCC, Y-Means and UAD. In FCC a new concept called fuzzy connectedness [10] is introduced. It is used to calculate the similarities among different data instances. Starting with a single or a few seed points in each cluster, all the data points are dynamically assigned to the cluster that has the highest fuzzy connectedness value (strongest connection). Comparing with the frequently utilized Euclidean distance, the new similarity metric is more robust and there is no restriction on the shape of clusters that can be discovered. FCC has also some other advantages over the other previously proposed clustering algorithms: no predefined “cluster width”, no threshold for “confidence area”, etc. Moreover, this unsupervised learning method can detect not only the known intrusion types, but also their variants. UAD and Y-Means methods detect intrusions based on feature vectors collected from the network, without being given any information about classifications of these vectors. There are two primary advantages of this system over signature-based classifiers or learning algorithms that require labelled data in their training sets. The first is that no manual classification of training data needs to be done. The second is that we do not have awareness about new types of intrusions for the system to be able to detect them. The main requirement is that the data conform to several assumptions. The systems then try to automatically determine which data instances fall in to the normal class and which ones are intrusions. Since no prior classification is required on the training data, and no knowledge is needed about new attacks, we can automate the process of training and create new cluster sets. In addition to that the method can be used for semi-automated detection by helping analysts to focus on portions of the data that are more likely to contain intrusions. From the result of the research comparing with the other two clustering algorithms such as UAD method and Y-Means the Detection Rate of FCC is at least 5% higher. Even though the False Alarm Rate with FCC is also a little higher, these results are very encouraging since higher Detection Rate is more important in most applications.

## 9. References

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