

# Intrusion Detection System Using Neural Networks

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## Overview

- Introduction
- Several Methods to detect intrusion
- A model of Intrusion detection System(IDS) using artificial neural networks(ANN)
- Analysis of IDS using ANN.
- Conclusion

## Introduction

- What is Intrusion ?  
Intrusion refers to any unauthorized access, unauthorized attempt to access or damage, or malicious use of information resources.
- Necessity Intrusion Detect Systems.
  1. With the growing rate of interconnections among computer systems, network security is becoming a real challenge.
  2. IDSs are designed to protect availability, confidentiality and integrity of critical networked information systems.

## Type of IDS

- Early research into IDS suggested two major detection principles:
- Anomaly Detection  
Attempts to quantify the usual or acceptable behavior and flags other irregular behavior as potentially intrusive.
  - Misuse Detection  
Attempts to flag behavior that is close to some previously defined pattern signature of a known intrusion.

## Methods to detect intrusion

### 1. Expert System

- Most common form of rule-based intrusion detection approach.
- Rule-based analysis relies on sets of predefined rules that are provided by an administrator, automatically created by the system, or both.
- Requires frequent updates to remain current.
- Less flexibility in the rule-to-audit record representation.

### 2. Statistical Method

- Often used to measure how anomalous the behavior is.
- Requires that the distribution of subject's behavior is known.
- Pattern matching techniques are used to determine whether the sequence of events is part of normal behavior.

## Methods to detect intrusion

### 3. Artificial Neural Networks

- ANNs have recently been proposed as alternatives to the statistical analysis components of anomaly detection system.
- Intrusion detection using neural networks is more flexible.
- The inherent ability of neural networks can learn the characteristic of attacks.

## Example: Real Time detection(NSOM)

### Model:

A Real-Time Network-based Intrusion Detection System Using Self-Organizing Maps.

### Hypothesis:

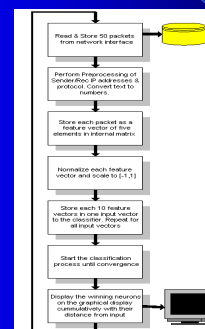
- The routine traffic that represents normal behavior would be clustered around one or more cluster centers ;
- Any irregular traffic representing abnormal and possibly suspicious behavior would be clustered outside of the normal clustering.

### Data collection:

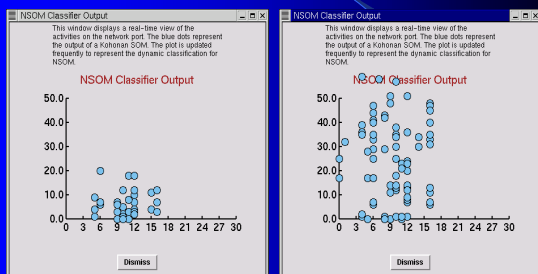
Use a host PC running Linux as the primary test bed. The system is connected to a network using Ethernet controller.

## NSOM

- Data processing:



## NSOM(result)



## NSOM(result)

- Compare Y value on the graphs:
  1. The attack neurons were much higher than normal ones.
  2. Since Y values represent the distance of the winning neurons with respect to the input vector, we can conclude high Y value neurons represent uncommon and irregular behavior and therefore a possible attack.

## Analysis

- This model shows that we were able to classify simulated DoS(Denial of service) attacks graphically as opposed to normal traffic, by showing different clustering of output neurons.
- Training requires more data, which may need more disk space and CPU resource.
- In order to analyze system log, the operating system must keep all performed action information.

## Conclusion

- Introduced the background of intrusion detection systems.
- Analyzed existing methods of IDS
- Intrusion detection using neural networks displayed substantial promise and it especially suited for multi-category classification.

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