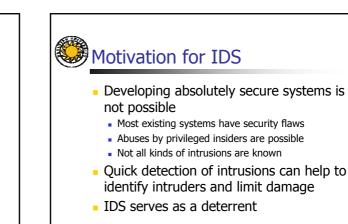
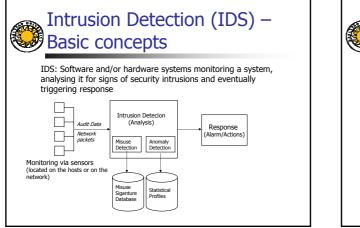
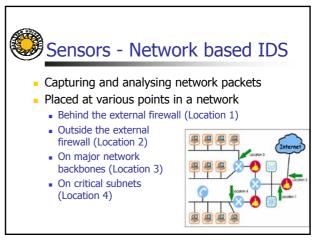
## Intrusion Detection (IDS)

Simone Fischer-Hübner

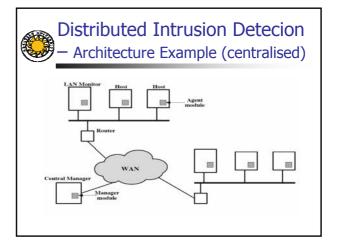






## Sensors – Host based IDS

- Information is collected within an individual computer or application
- Installed on critical hosts
- Audit data are collected on OS level (system logs) and/or application level



### Distributed Intrusion Detection – Distric Issues

- A distributed intrusion detection system may need to deal with different audit record formats
- One or more nodes in the network will serve as collection and analysis points for the data, which must be securely transmitted to them Architecture can be:
- centralized (single point of analysis, easier but bottleneck) or
- decentralized (multiple centers that must be coordinated)

# Anomaly Detection

- Based on the hypothesis that intrusions can be detected by monitoring a system for abnormal patterns of system usage
- **IDES** (Intrusion Detection Expert System) developed by D.Denning at SRI/International in 1986
- Usually rule-based pattern matching system which includés
  - Statistical profiles for representing the behavior of subjects with respect to objects
  - Rules matching new audit records against profiles, acquire/update profiles, detect anonalous behavior



- Attempted break-ins:
  - abnormally high rate of password failure
- Masquerading, successful break-ins different login time, location or connection type, different accesses to data, execution of programs
- Penetration by legitimate users:
- login at unusual times,route data to remote printers not normally used,
- execution of different programs, more protection violations, access to commands/files not normally permitted to him/her
- Viruses:
  - infected program needs more memory, disk space, CPU-time, I/O-activities,
  - it modifies other executable code not normally done by it,
  - increase in the frequency of executable files rewritten in the infected
  - system



Generated by the target system, translated into standard format, transmitted to the IDES system for analysis

<u>Audit record structure:</u> ( subject, action, object, exception-condition, resource-usage, time-stamp )

Decomposition of activities involving multiple objects to single-object actions: e.g.: COPY GAME.EXE to <LIBRARY>GAME.EXE issued by Smith is aborted, because he does not have write-permission to <LIBRARY>

Audit Records:

(Smith, execute, <Library>COPY.EXE, 0, CPU=0002, 1105821678) (Smith, read, <Smith>GAME.EXE, 0, RECORDS=0, 1105821679) (Smith, write, <Library>GAME.EXE, write-viol, Records=0, 1105821679)

### IDES Anomaly Detection -Statistical Profiles (I)

Profiles characterize the behaviour of a subject with respect to an object in terms of a *statistical metric* and *model* 

Metric

Random variable x representing a quantitative measure accumulated over a period (period: fixed or time between 2 events)

Examples of types of metrics: Event counter:

- x is the number of audit records satisfying some property occurring during a period, e.g. number of logins during one hour, number of execution failures during one session
- Interval timer.
- x is the length of time between two related events, e.g. time length between successive logins into one account Resource measure:
- x is the quantity of resources consumed by some action during a period, e.g. number of pages printed per day

### **IDES Anomaly Detection –** Statistical Profiles (II) Statistical Model: Given a metric for a random variable x and n observations x<sub>1</sub>,...,x<sub>n</sub> The statistical model shall determine whether a new observation $\boldsymbol{x}_{n+1}$ is abnormal with respect to the previous observations. **Operational Model:** Abnormality is detected by comparing a new observation of x against fixed limits, e.g. limitation of number of password failures during a short period.

Mean and Standard Deviation Model: A new observation of x is defined to be abnormal, if it falls outside a confidence interval: mean  $\pm$  d \* stdev (the probability of a value falling outside this interval is at most 1/d2).

# $sum = x_1 + x_2 + \dots + x_n$ sumsquares = $x_1^2 + \dots + x_n^2$

### mean = sum / n stdev = $\sqrt{}$ (sumsquares / (n-1) - mean<sup>2</sup>)

# IDES Example Profile

#### Profile structure:

(name, subject-pattern, action-pattern, object-pattern, exception-pattern, resource-usage-pattern, period, metric, statistical-model, value, threshold)

### Example of patterns:

Subject patterns: 'Smith', \*  $\rightarrow$  user Object patterns: '<Library>\*', IN(GAME.EXE,EDITOR.EXE)

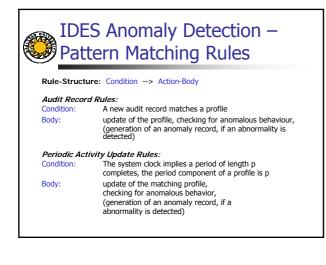


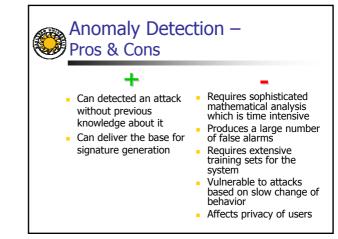
### Anomaly Detection – Examples of Metric/Model Combinations in Profiles

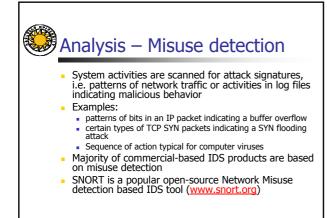
Login Frequency (event counter, mean/ standard deviation model) Location Frequency (event counter, mean/ standard deviation model) Session Output (resource measure, mean/ standard deviation model) Password Fails (event counter, operational model) Execution Frequency (event counter, mean/standard deviation model) Execution Denied (event counter, operational model)

Read-, Write,- Delete-Frequency (event counter, mean/standard deviation model)

Read-, Write, Delete-Fails (event counter, operational model) File Resource Exhaustion (event counter, operational model)





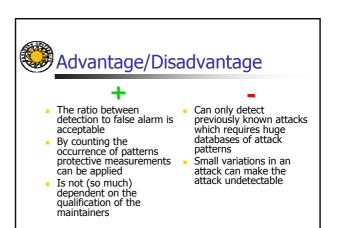


### Misuse Detection-Example: Buffer overflow attack signatures

- An exec system call audit records for a buffer overflow has the following pattern:
  - The exec call concerns a setuid program, i.e. the effective user id and the real user id fields are different
  - The argument passed to the exec call is relatively long, making the length of the entrire audit record significantly exceed the length al almost all normal setuid exec call
    - Buffer overflow atatcks typcially produce exec audit records with a length > 500 bytes. Only 0.15 % of normal exec audit records are longer than 400 bytes.
  - The exec argument contain opcode in the range of ascii control characters

Misuse Detection – Example: Virus signature					
	PSP J Program Before Infection				
	PSP J Program Virus J After Infection				
	cial Attack signature of com-Infectors (sequence of system calls): Open executable (.com) file to be infected Get date of last modification				
	Get time of last modification Read first 3 bytes to get jump address Go to end of file				
	Append code (of the virus) Go to beginning of the file				
	Write new jump address (3 bytes)				

- Reset date and time
- Close file



- Honeypots decoy systems to lure attackers away from accessing critical systems to collect information of their activities to encourage attacker to stay on system so administrator can respond are filled with fabricated information instrumented to collect detailed information on attackers activities
  - single or multiple networked systems
  - of IETF Intrusion Detection WG standards

## Anomaly Detection – Examples for Statistical Models (I)

Measure	Model	Type of Intrusion Detected			
Login and Session Activity					
Login frequency by day and time	Mean and standard deviation	Intruders may be likely to log in during off-hours.			
Frequency of login at different locations	Mean and standard deviation	Intruders may log in from a location that a particular user rarely or never uses.			
Time since last login	Operational	Break-in on a "dead" account.			
Elapsed time per session	Mean and standard deviation	Significant deviations might indicate masquerader.			
Quantity of output to location	Mean and standard deviation	Excessive amounts of data transmitted to remote location could signify leakage of sensitive data.			
Session resource utilization	Mean and standard deviation	Unusual processor or I/O levels could signal an intruder			
Password failures at login	Operational	Attempted break-in by password guessing.			
Failures to login from specified terminals	Operational	Attempted break-in.			

## Anomaly Detection – Examples for Statistical Models (II)

Execution frequency	Mean and standard deviation	May detect intruders, who are likely to use different commands, or a successful penetration by a legitimate user who has gained access to privileged commands.
Program resource utilization	Mean and standard deviation	An abnormal value might suggest injection of a virus or Trojan horse, which performs side-effects that increase I/O or processor utilization.
Execution denials	Operational model	May detect penetration attempt by individual user who seeks higher privileges.



File access activity					
Read, write, create, delete frequency	Mean and standard deviation	Abnormalities for read and write access for individual users may signify masquerading or browsing.			
Records read, written	Mean and standard deviation	Abnormality could signify an attempt to obtain sensitive data by inference and aggregation.			
Failure count for read, write, create, delete	Operational	May detect users who persistently attempt to access unauthorized files.			