

Protection of Communication Infrastructures Chapter 7 Intrusion Detection Systems

- Motivation
- Goals and Tasks of an IDS
- NIDS types & properties
- Intrusion Prevention
- Evading IDS

(Acknowledgement: some of slides have been adapted from [CDS05, Kön03])

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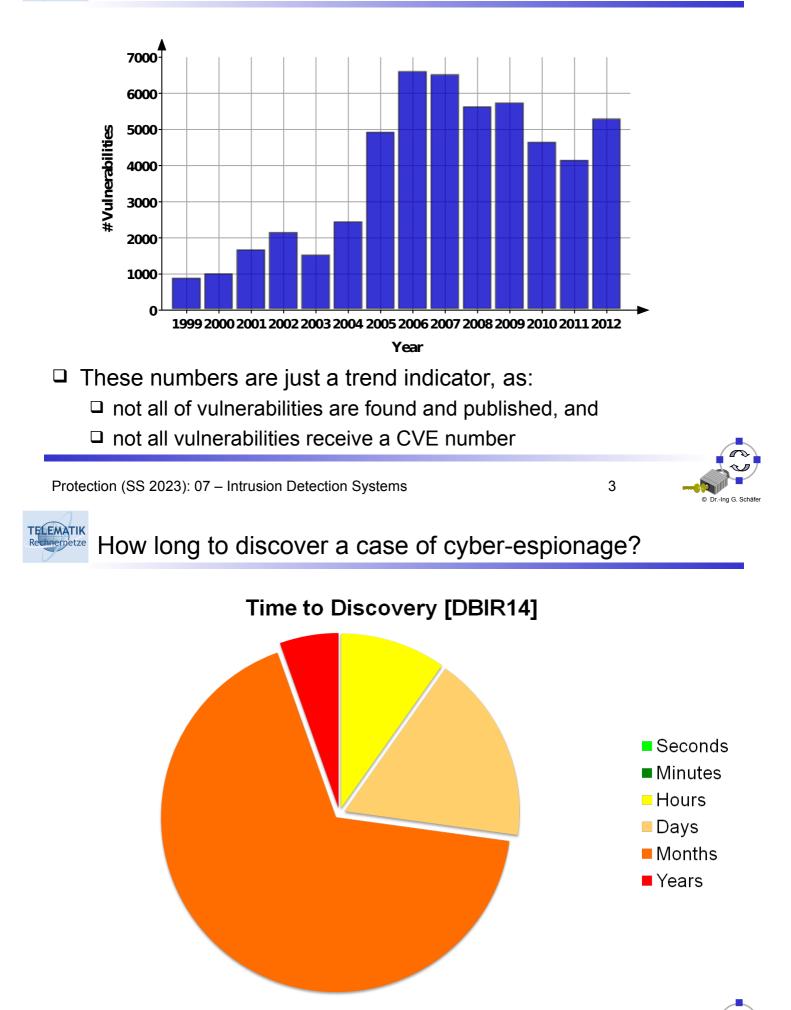


Introduction

- Definition:
 - An *intrusion* is an action or set of actions aimed at compromising the confidentiality, integrity or availability of a service or system
- Principal defense categories:
 - Prevention
 - Detection
 - □ Response

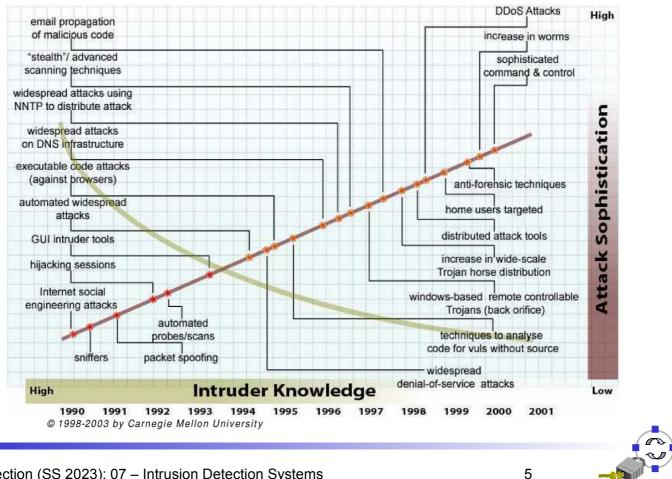


Number of vulnerabilities reported per year (CVE)





Attack Sophistication vs. Intruder Knowledge

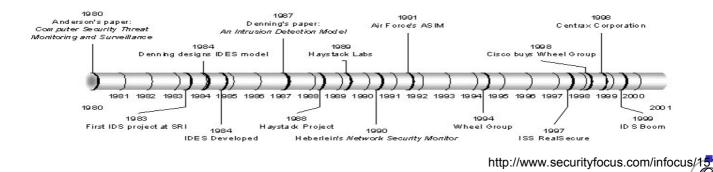


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TELEMATIK Rechnernetze

A Long History of Intrusion Detection

- 1980 James Anderson: Computer Security Threat Monitoring and Surveillance
- 1983 Dorothy Denning (SRI-International): Analysis of audit trails from government mainframe computers
- 1984 Dorothy Denning: Intrusion Detection Expert System (IDES)
- 1988 Lawrence Liverpool Laboratories: Haystack Projekt
- 1990 Heberlein: A Network Security Monitor (NSM)
- 1994 Wheel Group: First commercial NIDS (NetRanger)
- 1997 ISS: Real Secure
- Early 2000 Boom of Intrusion Detection System



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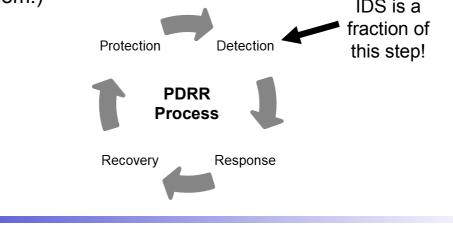
Goal of Intrusion Detection Systems

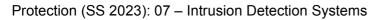
- Overall goal: Supervision of computer systems and communication infrastructures in order to detect intrusions and misuse
- Why detection of attackers?
 - □ Full protection is usually not possible!
 - Security measures too expensive or with too low flexibility, e.g., not possible to build every functionality in ASICs
 - □ Wrong postulates about capabilities of attackers (NSA?)
 - Unpatched systems for compliance reasons (medical systems etc.)
 - Because legitimate users get annoyed by too many preventive measures and may even start to circumvent them (introducing new vulnerabilities)
 - Because preventive measures may fail:
 - Incomplete or erroneous specification / implementation / configuration
 - Inadequate deployment by users (just think of passwords...)
- What can be attained with intrusion detection?
 - Detection of attacks and attackers
 - Detection of system misuse (includes misuse by legitimate users)

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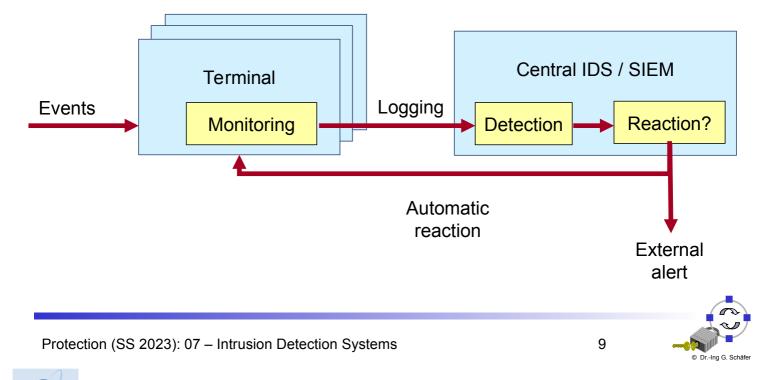
- Using a detection system only makes sense if there are consequences!
- Possible goals
 - Limitation of damage if (automated) response mechanisms exist
 - Gain of experience in order to recover from attack and improve preventive measures
 - Deterrence of other potential attackers (if and only if police is able to arrest them!)
 IDS is a









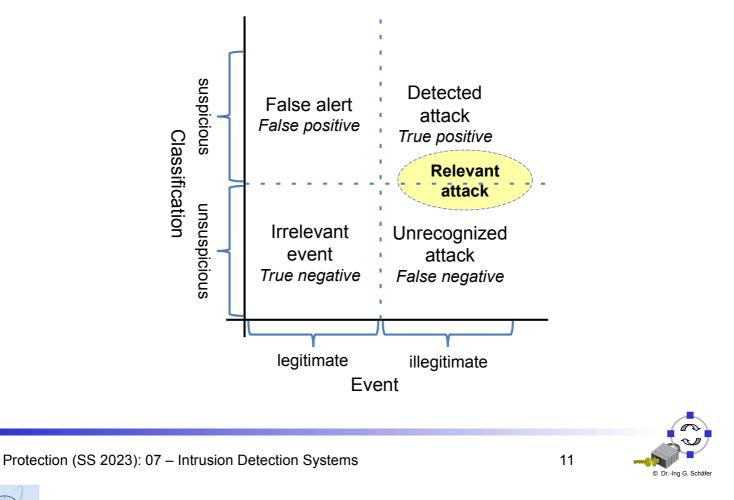


Tasks of an Intrusion Detection System

- □ Audit:
 - □ Recording of all security relevant events of a supervised system
 - Preprocessing and management of recorded audit data
- Detection:
 - Automatic analysis of audit data
 - Principle Approaches:
 - Signature analysis
 - Abnormal behavior detection (based on knowledge)
 - Anomaly detection (based on learned "normal level")
 - □ Types of errors:
 - False positive: a non-malicious action is reported as an intrusion
 - False negative: an intrusion is not detected (a "non-event")
- Response:
 - Reporting of detected attacks (alerts)
 - Potentially also initiating countermeasures (reaction)







Requirements to Intrusion Detection Systems

- □ High accuracy (= low rate of false positives and false negatives)
- □ Easy to integrate into a system / network
- □ Easy to configure & maintain
- Autonomous and fault tolerant operation
- □ Low resource requirements
- Self protection, so that an IDS itself can not easily be deactivated by a deliberate attack (in order to conceal subsequent attacks)



Classification of Intrusion Detection Systems

- □ Classification of intrusion detection systems (IDS):
 - □ Scope:
 - Host-based: analysis of system events
 - Network-based: analysis of exchanged information (IP packets)
 - Hybrid: combined analysis of system events and network traffic
 - □ Time of analysis:
 - Online analysis
 - Post mortem (Forensic tools, not covered here)



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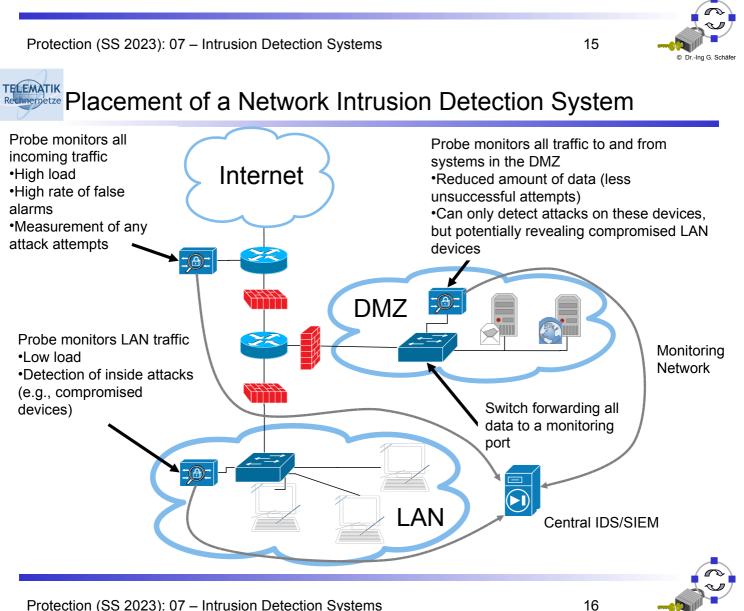


Host Intrusion Detection Systems (HIDS)

- □ Works on information available on a system:
 - OS and application logs
 - System file modification
 - Illegal file access
 - □ Login behavior (invalid tries, times)
 - Analysis of system resource consumption
 - □ Searches for viruses, rootkits etc.
- □ Can detect attacks by insiders (e.g. copy to USB stick), but:
 - Has to be installed on every system
 - Hard to manage on a large number of systems
 - Not available for every platform (e.g. routers, printers, medical devices etc.)
 - May be disabled by the attacker!
 - Produces lots of (potentially non-useful) information
 - □ Often no real-time analysis but predefined time intervals

TELEMATIK Network Intrusion Detection System (NIDS) Rechnernetze

- □ Analysis of network monitoring information (mostly on network layer)
- □ Existing systems use a combination of
 - Signature-based detection
 - Deviation from defined protocol behavior (stateful)
 - Statistical anomaly analysis
- □ Can even detect DoS with buffer overflow attacks, invalid packets, attacks on application layer, DDoS, spoofing attacks, port scans
- Often used on network hubs, to monitor a segment of the network \rightarrow Easier to manage & ensure monitoring of all devices
- □ (Obviously) cannot detect offline attacks, e.g., copy to USB stick
- □ In reality also produces lots of (potentially non-useful) information
- □ What about encrypted protocols?
- □ We will concentrate on NIDS in the following...



Intrusion Detection Message Exchange Format (1)

- Intrusion Detection Message Exchange Format (IDMEF)
 - IETF Intrusion Detection WG
 - □ RFC 4765 (Experimental)
 - Defines messages between probes and central components
 - □ Allows (in principle) to combine devices of different vendors
 - Object-oriented approach
 - XML-based encoding

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Intrusion Detection Message Exchange Format (2)

Message types

Heartbeat message

□ Alert message (ToolAlert, OverflowAlert, CorrelationAlert)

□ ...

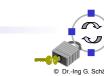
□ Event report

- Analyzer entity which emitted the alert
- Classification what attack has been detected
- Source any combination of multiple objects describing a network node, an user, a process, or a service
- Target any combination of multiple objects describing a network node, an user, a process, a service, or a file
- Assessment severity of the attack and confidence of the analyzer about the validity of the alert
- Additional information in (name, value) pairs



Recommerced Signature-based detection

- Basic idea:
 - $\hfill\square$ Some attack patterns can be described with sufficient detail \rightarrow specification of "attack signatures"
 - Event generated if packet(s) contains known attack signatures
- Identifying attack signatures:
 - Analyzing vulnerabilities
 - Analyzing past attacks that have been recorded in the audit
- Specifying attack signatures:
 - Based on identified knowledge so-called rules describing attacks are specified
 - Most IDS offer specification techniques for describing rules
 - Achievable detection quality directly dependent on quality of signature database (DB)



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Signature-based detection – Example: Snort (1)

□ Each detected attack type needs a predefined rule

```
alert icmp $EXTERNAL_NET any -> $HOME_NET any
  (msg: "Ping-of-Death detected";
   dsize: > 10000;
   sid: 3737844653)
```

- Shall detect Ping-of-Death packets, i.e., packets that are unusually large and crash the operating system
- □ How do these packets look in layer 3 (and below)
 - MTU is usually 1,500 bytes
 - $\Box \rightarrow$ at least 7 packets!
- □ Requires preprocessing of packets!





More sophisticated example, checking for mail server buffer overflows:

alert tcp \$EXTERNAL_NET any -> \$SMTP_SERVERS 25
(msg:"SERVER-MAIL RCPT TO overflow";
flow:to_server,established; Quick check
content:"rcpt to|3A|"; Better check
nocase;
isdataat:256,relative;
pcre:"/^RCPT TO\x3a\s*\x3c?[^\n\x3e]{256}/im";
classtype:attempted-admin;
sid:654;
rev:23;)
Very slow reqular
expression check



Signature-based detection – Packet Processing

- □ Three step processing of captured packets:
 - Preprocessing:
 - Normalized and reassembled packets (layer 3)
 - Recovery of TCP data flows (layer 4)
 - Normalization of application layer protocols
 - Detection engine works on the data and decides what action should be taken
 - Action is taken (log, alert, pass)

Signature-based detection – Properties

- □ Advantages:
 - Easy to setup
 - □ In some environments acceptable false positive rate
- Drawbacks:
 - Requires prior knowledge of all potential attacks
 - Signature database requires continuous updating
 - Large databases, difficult to maintain
 - Large number of "special plugins" for attacks not to express with rule language, e.g., to detect port scans
 - □ High false negatives rate if signature DB not adapted or up-to-date
 - □ IP & TCP preprocessing requires significant resources
 - Possibility of bypassing:
 - Attackers being aware of a certain IDS may try to craft attacks that are not covered by any signature
 - May be tested offline!

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- Basic idea detect behavior that differs significantly from normal use:
- □ Users and systems have "normal" use pattern:
 - Activity pattern
 - Used protocols & protocol states
 - Accessed servers
 - □ Traffic volumes etc.
- □ Assumption: "behavior" can be described by an administrator
 - Needs a specification, e.g., in a rule language
 - $\hfill\square$ For generic protocols such a description may be predefined
- □ Analysis:
 - Events matched against rules
 - Any mavericks will be reported
 - □ Comparable to a firewall that only performs logging...



Detection of Abnormal Behavior – Example Systems

- □ NetSTAT [VK98]
 - Early academic example
 - Compares network traffic in probes with fact base
 - □ Simple application layer inspection, e.g., NFS
- □ StealthWatch (commercial)
 - Commercial system
 - □ Analyses flow information in switches (e.g. Cisco NetFlow or sFlow)
 - Can detect network scans, worm spreading, DoS attacks …
- Bro Security Monitor
 - Long-living open source project
 - Performs stateful protocol analysis
 - Reports protocol deviations, e.g., undocumented commands
- □ (Honey pots & honey networks)
 - Systems not accessed by legitimate users by design
 - □ All access may be considered illegitimate...

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Detection of Abnormal Behavior – Properties

- □ Advantages:
 - Approach can detect unknown attacks
 - Attacks cannot easily be prepared to be not detected
 - □ If well set up: acceptable false positive rate
 - Events rather easy to interpret
- Drawbacks:
 - High administrative effort
 - □ Some attacks (e.g. buffer overflows) are most likely not detected
- Direct firewall integration perhaps better...



Automatic Anomaly Detection – Overview

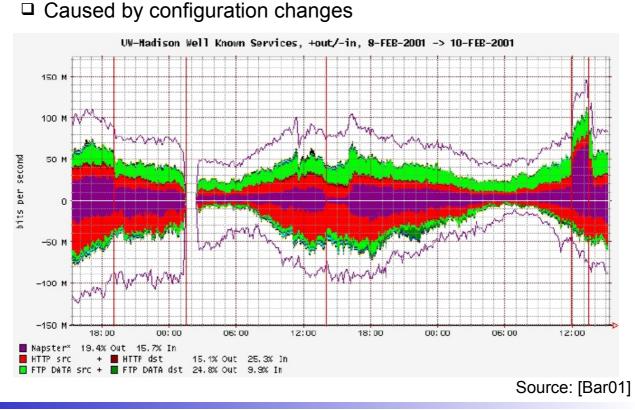
- Basic idea detect behavior that differs significantly from normal use, which is automatically learned
- Assumption: "normal user behavior" can be described statistically
 - Requires a learning phase / specification of normal behavior
 - Can learn significantly more features than an administrator is able to specify manually!
- □ Analysis:
 - Compares recorded events with reference profile of normal behavior
 - Use statistics and anomaly detection techniques to find outliers
 - Report if there is a timely correlation of a significant number of outliers



Recharged Ze Automatic Anomaly Detection - Example

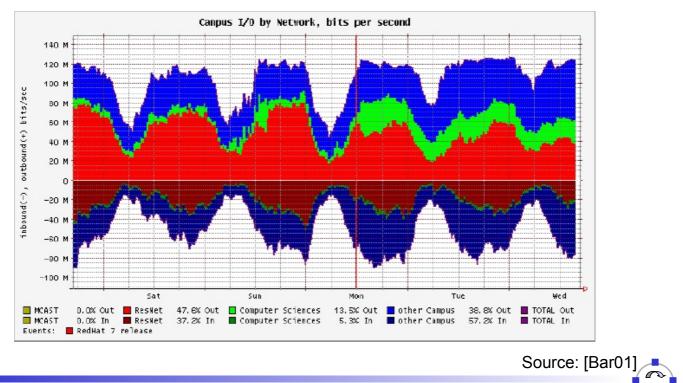
Automatic Anomaly Detection – Example (1)

□ Network operation anomalies



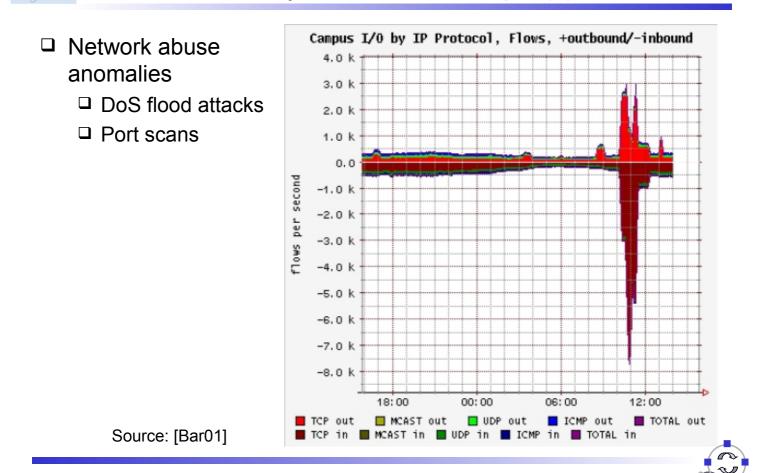
"Flash crowd anomalies"

Caused by software releases or special interest in a web site



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Recompletze Automatic Anomaly Detection – Example (3)



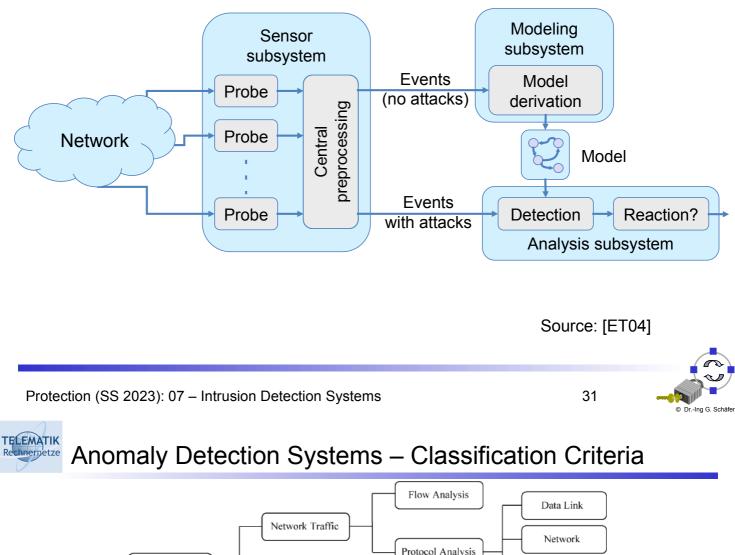
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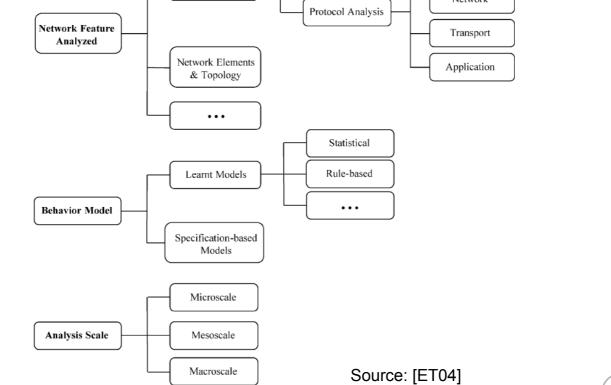
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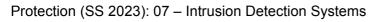
Generic anomaly detection system





Automatic Anomaly Detection – Anomaly Types [CBK09]

- Point Anomalies
 - □ Measurement points in an n-dimensional space (the lower the better → curse of dimensionality)
 - □ "Lonely" points or points of a small group are outliers
- Contextual Anomalies
 - Data points that are themselves not suspicious, but in their context
 - Example: Large data transfers from embedded device, low traffic at peak time
- Collective Anomalies
 - Detect deviations from a state machine
 - Data points are unsuspicious as long as they happen in a certain order
 - Deviations will be threated as an anomaly
 - Examples:
 - Retrieval of files without previously successful login (new state transition)
 - Usage of previously unused IP addresses (new state)





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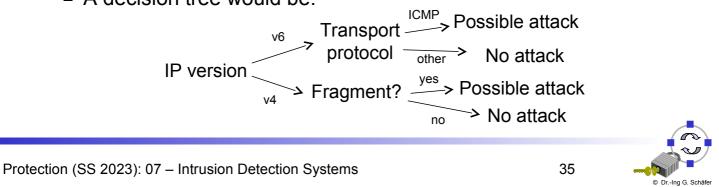
Automatic Anomaly Detection – Detection Methods [CBK09]

- Statistical Profiling
 - "Simple" statistical means, e.g., generating histograms, estimate parameters of distributions by maximum likelihood estimations, use regression methods to estimate curve parameters
 - □ Any significant change \rightarrow alert
- Neural Networks
 - Neuronal networks learn normal behavior and are trained to detect attacks
 - Different designs possible, e.g., Self-Organizing Maps (SOM) to detect outliers
- Bayesian Networks
 - Method developed for artificial intelligence
 - Events are nodes in a graph, edges model dependence
 - Probabilities and dependencies are learned automatically
 - □ System concludes using packet information, e.g., there are only few attacks for IPv6 and few attacks use small packets → small IPv6 packets are o.k.!



Automatic Anomaly Detection – Detection Methods [CBK09]

- Support Vector Machines
 - Finding functions that separate data points caused by different machines, i.e., data points from compromised/uncompromised devices
 - Other machines also in the space of the compromised machines might also be compromised
- Rule-based Learning
 - □ Automatic learning of rules to sort out anomalies, e.g., decision trees
 - □ Example:
 - Consider there are only ICMP-based attacks for IPv6 and fragment-based attacks for IPv4
 - A decision tree would be:





- Clustering-based
 - Measured data points may be separated into clusters
 - If attacks are more seldom than legitimate traffic (as it should be) smaller clusters are classified to be malicious
 - Generally resource-intensive to calculate (NP-hard)
 - Popular approximation: k-Means
- Nearest-Neighbor-based
 - Simple alternative to clustering: calculate distance to closest neighbors
 - High distances indicate outliers
- Information-theory-based
 - □ Calculate information theoretic metrics for the normal traffic, e.g., entropy
 - When there are new traffic patterns (what could be attacks) entropy increases
 - Example: Compression of HTTP requests, if there is shell-code in it, it should be different from previous requests and less compressible



- □ Spectral analysis
 - Actually two methods
 - □ In time-series:
 - Derive patterns of recurring values, e.g., large file transfers once a month for backups are ok
 - E.g. using Fourier transformation
 - □ In graphs:
 - Reduction of dimensionality of large matrixes
 - Example: Calculation of eigenvalues in an adjacency matrix, modeling the devices communicating with each other
 - Spectral gap (difference between the two largest eigenvalues) indicates connectivity of the graph



Automatic Anomaly Detection – Example: PHAD

- Packet Header Anomaly Detection (PHAD) [Mah01]
- □ Old academic example, but comparably good results (back then)
- Simple protocol analysis, "learns" normal ranges of values for each header field (link, network, transport layer)
- Other values are classified anomalous

$$score_{packet} = \sum_{i \in \text{ anomalous fields}} \frac{t_i n_i}{r_i}$$

$$t \dots \quad \text{time since previous anomaly}$$

$$n \dots \quad \text{number of observations}$$

$$r \dots \quad \text{number of distinct values}$$

Learning phase + detection phase



Automatic Anomaly Detection – Example: ALAD

- Application Layer Anomaly Detection (ALAD) [Mah02]
- □ Extension to PHAD, introduces conditional probabilities
- □ Five models:

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□ P(src IP | dest IP)

Learns normal set of clients for each host, i.e., the set of clients allowed on a restricted service

- □ P(src IP | dest IP, dest port)
 - Like (1), but one model for each server on each host
- P(dest IP, dest port)

Learns the set of local servers which normally receive requests

- P(TCP flags | dest port)
 Learns the set of TCP flags for all packets of a particular connection
- □ P(keyword | dest port)

Examines the text in the incoming request (first 1000 bytes)

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Automatic Anomaly Detection – Properties

- □ Advantages:
 - Can detect unknown attacks
 - Comparably easy to setup

Drawbacks:

- Privacy:
 - Collecting user specific usage patterns
 - Work-related or personal habits
- Requires continuous refreshing of normal behavior patterns
- High number of false positives
- Even true positives often difficult to interpret
- □ If a normal behavior pattern matches an attack pattern, this kind of attack will not be detected (→ false negative)
 - What about the regular refreshes of the model?



Testing and Benchmarking of IDS

- DARPA Environment (1998/1999)
 - First systematic effort to test an IDS
 - □ Analysis of huge amounts of data, e.g. from Hanscom Air Force Base
- □ LARIAT Environment (2000)
 - Lincoln Adaptive Real-time Information Assurance Test-bed
 - Emulates network traffic from a small organization
 - □ Traffic generation using defined service models
- □ Predominant open source philosophy for testing an IDS
 - Individual test environment
 - Search for existing exploits / attacks
 - □ Mix of background traffic and attack traffic
 - □ Analysis of the detection ratio (false positive / false negative)



Summary: Properties of the approaches

- □ Signature-based Detection:
 - Requires high effort in specification of rules (can be leveraged by multiple usage; comparable to sharing of virus description)
 - □ Effective detection of attacks that have been described in rule database
 - Unknown attacks cannot be detected
- Detection of Abnormal Behavior
 - Extremely high effort to set up
 - Possibility to detect some unknown attacks
- □ Anomaly Detection:
 - Theoretically challenging
 - Realization expensive in terms of required data and analysis capabilities
 - Limited Effectiveness

 Approaches represent complementary techniques (rather than antagonistic ones)

Intrusion Prevention Systems – Motivation

- Automatic event generation nowadays not sufficient
 - □ Automatic exploitation is extremely fast → human intervention would be too late
 - □ Too many attacks on current systems → must be handled automatically for reasons of efficiency
- □ Led to the development of Intrusion Prevention Systems (IPS)
- Differentiation between IDS and IPS no longer meaningful as nearly all modern IDS are also IPS



Intrusion Prevention Systems – Approaches (1)

- □ Inline operation and suppression
 - □ All traffic is going through the IPS
 - Any flow (and possibly similar flows) generating an attack event will be suppressed
 - □ Pros:
 - Efficient
 - No race conditions
 - Cons:
 - Possible bottleneck and single point of failure
 - May be difficult to set up

Intrusion Prevention Systems – Approaches (2)

- □ Firewall reconfiguration
 - IPS reconfigures an existing firewall to suppress suspicious flows
 - Pros:
 - Relatively easy to set up
 - No single points of failure
 - Cons:
 - Race conditions (what if the attack already reached the target, especially if the IPS is under load?)
- Sending TCP-RST packets
 - IPS resets TCP flows by resetting the connection
 - Pros:
 - Extremely easy to setup
 - No single point of failure
 - Cons:
 - Race conditions
 - Works only for TCP

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Intrusion Prevention Systems – Approaches (3)

- Deflection
 - Reconfiguration of firewall and/or routers
 - Attacker is transparently redirected to honey pots to slow down his attack
 - Pro:
 - May cause a significant slow down / confusions
 - Cons:
 - Difficult to setup (if done well)
 - Race conditions?!
- Active Defense or Automatic Hack-back
 - Academic approach (fortunately)
 - Attacks cause a manual or automatic "strike-back"
 - Used already in early 1990s by the US military to unveil "stepping stones", i.e., proxies used by an attacker to protect his identity



- □ Using IPS may be an option...
 - Realized approach depends on scenario
 - Not a replacement for fixing software!
- □ Always requires a detailed risk analysis:
 - Will the damage caused by false positives and the automatic suppression of legitimate flows, be lower than the damage prevented by suppression of illegitimate flows?
 - □ What about attacks from spoofed IP addresses?
- Usually only suitable for closed, well-controlled network environments...

□ E.g. preventing SQL injections in a web server

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- □ Anomaly detection:
 - □ Attacker may act slowly
 - May generate high amount of "legitimate traffic" to cover attack
 ...
- Signature-based IDS: Attackers may try to construct attacks such that they are not detected
 - Works extremely well when the attacker has access to the rule set
 - May even be automated...
 - Requires countermeasures in IDS (sometimes extremely complicated)



- Popular methods to obfuscate attacks:
 - Recode URLs
 - Characters in URL may be expressed by different encodings
 - Example: 'a', '%61' and '%u0061' express all the same letter
 - Relatively easy to revert, but requires TCP reassembly
 - Recode shell code
 - Encrypt parts of the shell code (and decrypt on the fly)
 - Use different commands to achieve the same thing
 - Insert dummy commands to change the signature
 - Example: Change NOP slide from 0x90 0x90 0x90 0x90 0x90 0x90 to 0x0c0c 0x0c0c 0x0c0c (3 times decrease register AH by 12)
 - Extremely difficult to revert

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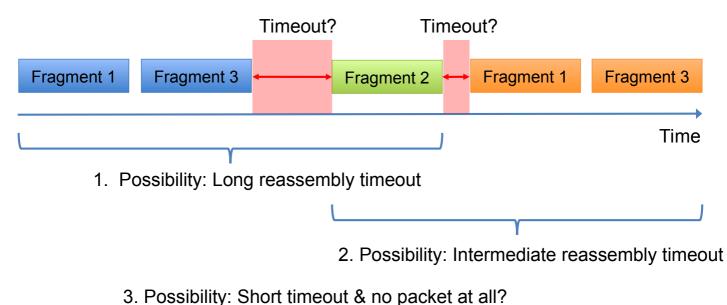
IDS Evasion – Constructing Packets

- Observation: Packet processing in IDS & end-system must be the same (otherwise different PDUs are reconstructed)
- Problem: Different OSes treat packets different as standards are ambiguous
- □ Examples: Overlapping TCP segments and IP fragments
 - □ Some OSes use first PDU part others the last send one etc.
 - IDS must either know the OS of the end-system or try all possible combinations
- Even more problematic: IDS may see packets that the end-system does not
 - Example: 1. Attacker sends (legal) TCP flow, 2. He sends a single TCP RST packet with a TTL s.t. a router behind the IDS drops it, Attacker continues TCP flow with exploit, while IDS believes in out of order packets



IDS Evasion – Considering timeouts

- Most problematic: Timeouts depend on OS & delays (especially jitter)
- Example: Timeouts for IP reassembly



Cannot be decided securely!
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General Problems of IDS (1)

- Audit Data:
 - Amount of log data:
 - Auditing often generates a rather high data volume
 - \Rightarrow Significant storage capacities are required
 - \Rightarrow Processing of audit data should be automated as much as possible
 - Location of audit data storage:
 - Alternatives: on specific "log server" or the system to be supervised
 - \Rightarrow If stored on log server, data must be transferred to this server
 - \Rightarrow If stored on the system to be supervised, the log uses significant amounts of resources of the system
 - Protection of audit data:
 - If a system gets compromised, audit data stored on it might get compromised either
 - □ Expressiveness of audit data:
 - Which information is relevant?
 - Audits often contain a rather low percentage of useful information



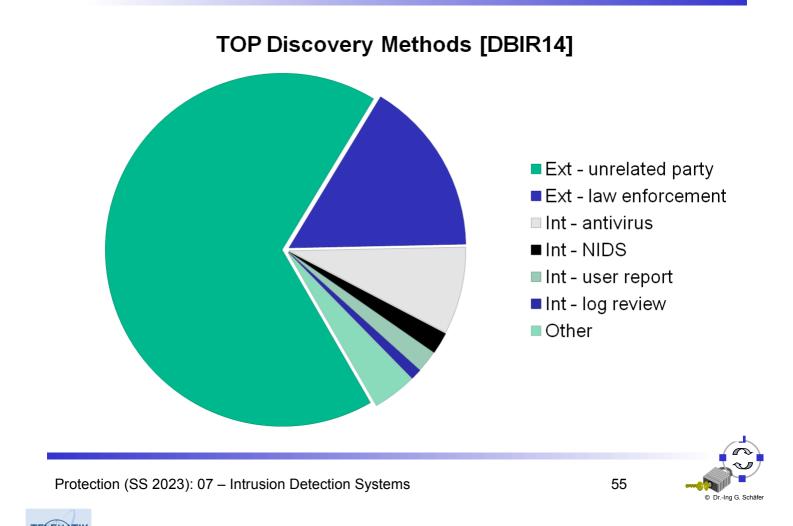
- \Box Privacy (\rightarrow "Datenschutz"):
 - □ User identifying data elements are logged, e.g.:
 - Directly identifying elements: user ids
 - Indirectly / partly identifying elements: names of directories and subdirectories (home directory), file names, program names
 - Minimally identifying elements: host type + time + action, access rights + time + action
 - □ IDS audits may violate the privacy of users:
 - Violation of the user's right to determine himself which data is collected regarding his person
 - Collected information might be abused if not secured properly
 - □ Potential solution:
 - Pseudonymous audit: log activities with user pseudonyms and ensure, that they can only be mapped to user ids upon incident detection

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General Problems of IDS (3)

- Limited efficiency of analysis:
 - Most IDS follow a centralist approach for analysis: so-called agents collect audit data and one central evaluation unit analyzes this data
 - \Rightarrow No (partial) evaluation in agents
 - \Rightarrow Performance bottleneck
 - Insufficient efficiency, especially concerning attack variants and attacks with parallel actions
- □ High number of false positives:
 - In practice, many IDS report too many false alarms (some publications report up to 10.000 per month)
 - □ Potential countermeasure: alarm correlation (→ hierarchical approach)
- □ Further problems / open issues:
 - Self protection (including strategies to cope with high load)
 - □ High maintenance overhead
 - Cooperation between multiple IDS

Reality check: How is cyber espionage discovered?



Additional References (1)

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