ABSTRACT
We introduce a new simulation platform called Insight, created to design and simulate cyber-attacks against large arbitrary target scenarios. Insight has surprisingly low hardware and configuration requirements, while making the simulation a realistic experience from the attacker’s standpoint. The scenarios include a crowd of simulated actors: network devices, hardware devices, software applications, protocols, users, etc.

A novel characteristic of this tool is to simulate vulnerabilities (including 0-days) and exploits, allowing an attacker to compromise machines and use them as pivoting stones to continue the attack. A user can test and modify complex scenarios, with several interconnected networks, where the attacker has no initial connectivity with the objective of the attack.

We give a concise description of this new technology, and its possible uses in the security research field, such as pentesting training, study of the impact of 0-days vulnerabilities, evaluation of security countermeasures, and risk assessment tool.

Categories and Subject Descriptors
I.6.7 [Simulation Support Systems]; I.6.3 [Simulation and modeling]: Applications

General Terms
Security, Experimentation

Keywords
network security, network simulation, penetration test, vulnerability, exploit, 0-day, cyber-attack, training

1. INTRODUCTION
Computer security has become a necessity in most of today’s computer uses and practices, however it is a wide topic and security issues can arise from almost everywhere: binary flaws (e.g., buffer overflows [17]), Web flaws (e.g., SQL injection, remote file inclusion), protocol flaws (e.g., TCP/IP flaws [3]), not to mention hardware, human, cryptographic and other well known flaws.

Although it may seem obvious, it is useless to secure a network with a hundred firewalls if the computers behind it are vulnerable to client-side attacks. The protection provided by an Intrusion Detection System (IDS) is worthless against new vulnerabilities and 0-day attacks. As networks have grown in size, they implement a wider variety of more complex configurations and include new devices (e.g. embedded devices) and technologies. This has created new flows of information and control, and therefore new attack vectors. As a result, the job of both black hat and white hat communities has become more difficult and challenging.

The previous examples are just the tip of the iceberg, computer security is a complex field and it has to be approached with a global view, considering the whole picture simultaneously: network devices, hardware devices, software applications, protocols, users, etcetera. With that goal in mind, we are going to introduce a new simulation platform called Insight, which has been created to design and simulate cyber-attacks against arbitrary target scenarios.

In practice, the simulation of complex networks requires to resolve the tension between the scalability and accuracy of the simulated subsystems, devices and data. This is a complex issue, and to find a satisfying solution for this trade-off we have adopted the following design restrictions:

1. Our goal is to have a simulator on a single desktop computer, running hundreds of simulated machines, with a simulated traffic realistic only from the attacker’s standpoint.
2. Attacks within the simulator are not launched by real attackers in the wild (e.g. script-kiddies, worms, black hats). As a consequence, the simulation does not have to handle exploiting details such as stack overflows or
heap overflows. Instead, attacks are executed from an attack framework by Insight users who know they are playing in a simulated environment.

To demonstrate our approach, Insight introduces a platform for executing attack experiments and tools for constructing these attacks. By providing this ability, we show that its users are able to design and adapt attack-related technologies, and have better tests to assess their quality. Attacks are executed from an attack framework which includes many information gathering and exploitation modules. Modules can be scripted, modified or even added.

One of the major Insight features is the capability to simulate exploits. An exploit is a piece of code that attempts to compromise a computer system via a specific vulnerability. There are many ways to exploit security holes. If a computer programmer makes a programming mistake in a computer program, it is sometimes possible to circumvent security. Some common exploiting techniques are stack exploits, heap exploits, format string exploits, etc.

To simulate these techniques in detail is very expensive. The main problem is to maintain the complete state (e.g., memory, stack, heap, CPU registers) for every simulated machine. From the attacker’s point of view, an exploit can be modeled as a magic string sent to a target machine to unleash a hidden feature (e.g., reading files remotely) with a probabilistic result. This is a lightweight approach, and we have sacrificed some of the realism in order to support very large and complex scenarios. For example, 1,000 virtual machines and network devices (e.g., hubs, switches, IDS, firewalls) can be simulated on a single Windows desktop, each one running their own simulated OS, applications, vulnerabilities and file systems. Certainly, taking into account available technologies, it is not feasible to use a complete virtualization server (e.g., VMware) running thousands of images simultaneously.

As a result, the main design concept of our implementation is to focus on the attacker’s point of view, and to simulate on demand. In particular, the simulator only generates information as requested by the attacker. By performing this on-demand processing, the main performance bottleneck comes from the ability of the attacker to request information from the scenario. Therefore, it is not necessary, for example, to simulate the complete TCP/IP packet traffic over the network if nobody is requesting that information. A more lightweight approach is to send data between network sockets writing in the memory address space of the peer socket, and leaving the full packet simulation as an option.

2. BACKGROUND & RELATED WORK
Using simulated networks as a research tool to gather knowledge regarding the techniques, strategy and procedures of the black hat community is not a new issue. Solutions such as honeypots and honeynets [20, 25] were primarily designed to attract malicious network activities and to collect data.

A precise definition for the term honeypot is given by The Honeynet Project [19]:

A honeypot is an information system resource whose value lies in unauthorized or illicit use of that resource.

Over the last decade a wide variety of honeypot systems were built [15, 2, 24, 28, 21], both academic and commercial. Honeypots have emerged as an interesting tool for gathering knowledge on new methods used by attackers, and the underlying strength of the approach lies in its simplicity. Typically, honeypots offer minimal interaction with the attacker, emulating only small portions of a real network behavior. However, this simplicity is also a weakness: none of these systems execute kernel or application code that attackers seek to compromise, and only a few ones maintain a per-flow and per-protocol state to allow richer emulation capabilities. Thus, honeypots are most useful for capturing indiscriminate or large-scale attacks, such as worms, viruses or botnets, rather than very focused intrusions targeting a particular host [20].

In Table 1, we show the main differences with our approach. In particular, we are interested in the ability to compromise machines, and use them as pivoting stones to build complex multi-step attacks.

| Honeypots-like tools | Insight |
|----------------------|---------|
| Design focus: to detect, understand and monitor real cyber-attacks. | Design focus: to reproduce or mimic cyber-attacks, penetration test training, what-if and 0-day scenarios. |
| Attacks are launched by real attackers: worms, botnets, script-kiddies, black-hats. | Attacks are launched by the Insight users: pentest and forensic auditors, security researchers. |
| Simulation up to transport layer. | Simulation up to application layer, including vulnerabilities and exploits. |
| Stateless or (a kind of) per-flow and per-protocol state. | Applications and machines internal state. |
| No exploit simulation. No pivoting support. | Full exploit and agent (shellcode) simulation. Ability to pivot through a chain of agents. |

Table 1: Honeypots vs. Insight.

In contrast, “high interaction honeypots” and virtualization technologies (e.g., VMware, Xen, Qemu) execute native system and application code, but the price of this fidelity is quite high. For example, the RINSE approach [13] is implemented over the iSSFNet network simulator, which runs on parallel machines to support real-time simulation of large-scale networks. All these solutions share the same principle of simulating almost every aspect of a real machine or real network, but share the similar problems too: expensive configuration cost and expensive hardware and software licenses. Moreover, most of these solutions are not fully compatible with standard network protections (e.g., firewalls, 1

1 In a network attack, to pivot means to use a compromised machine as a stepping stone to reach further networks and machines, making use of its trust relationships.
IDSs), suffering a lack of integration between all security actors in complex cyber-attack scenarios.

Security assessment and staging are other well known security practices. It is common, for example in web application development, to duplicate the production environment on a staging environment (accurately mimicking or mirroring the first) to anticipate changes and their impact. The downside is: it is very difficult to adopt this approach in the case of network security due to several reasons. It would require the doubling of the hardware and software licenses and (among other reasons) there are no means to automatically configure the network.

Other interesting approaches to solve these problems include the framework developed by Bye et al. [4]. While they focus on distributed denial of service attacks (DDoS) and defensive IDS analysis, we focus on offensive strategies to understand the scenarios and develop countermeasures. Also Loddo et al. [14] have integrated User Mode Linux [9] and Virtual Distributed Ethernet [8] to create a flexible and very detailed network laboratory and simulation tool. The latter project has privileged accuracy and virtualization over scalability and performance.

The Potemkin Virtual Honeyfarm [20] is another interesting prototype. It improves high-fidelity honeypot scalability by up to six times while still closely emulating the execution behavior of individual Internet hosts. Potemkin uses quite sophisticated on-demand techniques for instantiating hosts [20], but this approach focuses on attracting real attacks and it shows the same honeypot limitations to reach this goal. As an example, to capture e-mail viruses, a honeypot must posses an e-mail address, must be scripted to read mail (executing attachments like a naive user) and, most critically, real e-mail users must be influenced to add the honeypot to their address books. Passive malware (e.g., many spyware applications) may require a honeypot to generate explicit requests, and focused malware (e.g., targeting only financial institutions) may carefully select its victims and never touch a large-scale honeyfarm. In each of these cases there are partial solutions, and they require careful engineering to truly mimic the target environment.

In conclusion, new trends in network technologies make cyber-attacks more difficult to understand, learn and reproduce, and the current tools to tackle these problems have some deficiencies when facing large complex scenarios. In spite of that, it is possible to overcome the problems described above using the lightweight software simulation tool we present.

3. INSIGHT APPROACH & OVERVIEW
A diagram of the Insight general architecture is showed in Fig. 1. The Simulator subsystem is the main component. It performs all simulation tasks on the simulated machines, such as system call execution, memory management, interrupts, device I/O management, etcetera.

At least one Simulator subsystem is required, but the architecture allows several ones, each running in a real computer (e.g., a Windows desktop). In this example, there are two simulation subsystems, but more could be added in order to support more virtual hosts.

The simulation proceeds in a lightweight fashion. It means, for example, that not all system calls for all OS are supported by the simulation. Instead of implementing the whole universe of system calls, Insight handles a reduced and generic set of system calls, shared by all the simulated OS. Using this approach, a specific OS system call is mapped to an Insight syscall which works similarly to the original one. For example, the Windows sockets API is based on the Berkeley sockets API model used in Berkeley UNIX, but both implementations are slightly different. Similarly, there are some instances where Insight sockets have to diverge from strict adherence to the Berkeley conventions, usually due to implementation difficulties in the simulated environment. In spite of this (and ignoring the differences between OS), all sockets system calls of the real world have been mapped to this unique simulated API.

Of course, there are some system calls and management tasks closely related to the underlying OS which were not fully supported, such as UNIX fork and signal syscalls, or the complete set of functions implemented by the Windows SDK. There is a trade-off between precision and efficiency, and the decision of which syscalls were implemented was made with the objective of maintaining the precision of the simulation from the attacker’s standpoint.

Figure 1: Insight architecture layout.

3For additional details look at the Winsock API documentation (available from http://msdn.microsoft.com), which includes a section called Porting Socket Applications to Winsock.
The exploitation of binary vulnerabilities is simulated with a probabilistic approach, keeping the attack model simple, lightweight, and avoiding to track anomalous conditions (and its countermeasures), such as buffer overflows, format string vulnerabilities, exception handler overwriting—among other well known vulnerabilities. This probabilistic approach allows us to mimic the unpredictable behavior when an exploit is launched against a targeted machine.

Let us assume that a simulated computer was initialized with an underlying vulnerability (e.g. it hosts a vulnerable OS). In this case, the exploit payload is replaced by a special ID or “magic string”, which is sent to the attacked application using a preexistent TCP communication channel. When the attacked application receives this ID, Insight will decide if the exploit worked or not based on a probability distribution that depends on the exploit and the properties describing the simulated computer (e.g., OS, patches, open services). If the exploit is successful, then Insight will grant the control in the target computer through the agent abstraction, which will be described in §4.

The probabilistic attack model is implemented by the Simulator subsystems, and it is supported by the Exploits Database, a special configuration file which stores the information related to the vulnerabilities. This file has a XML tree structure, and each entry has the whole necessary information needed by the simulator to compute the probabilistic behavior of a given simulated exploit. For example, a given exploit succeeds against a clean XP SP2 with 83% probability if port 21 is open, but crashes the system if it is a SP1. We are going to spend some time in the probability distribution, how to populate the exploits database, and the Insight attack model in the next sections.

Returning to the architecture layout showed in Fig. 1 all simulator subsystems are coordinated by a unique Simulator Monitor, which deals with management and administrative operations, including administrative tasks (such as starting/stopping a simulator instance) and providing statistical information for the usage and performance of these.

A set of Configuration Files defines the snapshot of a virtual Scenario. Similarly, a scenario snapshot defines the instantaneous status of the simulation, and involves a crowd of simulated actors: servers, workstations, applications, network devices (e.g. firewalls, routers or hubs) and their present status. Even users can be simulated using this approach, and this is especially interesting in client-side attack simulation, where we expect some careless users opening our poisoned crafted e-mails.

Finally, at the right bottom of the architecture diagram, we can see the Penetration Testing Framework, an external system which interacts with the simulated scenario in real time, sending system call requests through a communication channel implemented by the simulator. This attack framework is a free tailored version of the Impact solution however other attack tools are planned to be supported in the future (e.g., Metasploit).

The attacker actions are coded as Impact script files (using Python) called modules, which have been implemented using the attack framework SDK, as shown in the architecture diagram. The framework Python modules include several tools for common tasks (e.g. information gathering, exploits, import scenarios). The attacks are executed in real time against a given simulated scenario; a simulation component can provide scenarios of thousands of computers with arbitrary configurations and topologies. Insight users can design new scenarios and they have scripts to manage the creation and modifications for the simulated components, and therefore iterate, import and reproduce cyber-attack experiments.

4. THE SIMULATED ATTACK MODEL

One of the characteristics that distinguish the scenarios simulated by Insight is the ability to compromise machines, and use them as pivoting stones to build complex multi-step attacks. To compromise a machine means to install an agent that will be able to execute arbitrary system calls (syscalls) as a user of this system.

The agent architecture is based on the solution called syscall proxy (see §3 for more details). The idea of syscall proxying is to build a sort of universal payload that allows an attacker to execute any system call on a compromised host. By installing a small payload (a thin syscall server) on a vulnerable machine, the attacker will be able to execute complex applications on his local host, with all system calls executed remotely. This syscall server is called an agent.

In the Insight attack model, the use of syscall proxying introduces two additional layers between a process run by the attacker and the compromised OS. These layers are the syscall client layer and the syscall server layer.

The syscall client layer runs on the attacker’s Penetration Testing Framework. It acts as a link between the process running on the attacker’s machine and the system services on a remote host simulated by Insight. This layer is responsible for forwarding each syscall argument and generating a proper request that the agent can understand. It is also responsible for sending this request to the agent and sending back the results to the calling process.

The syscall server layer (i.e. the agent that runs on the simulated system) receives requests from the syscall client to execute specific syscalls using the OS services. After the syscall finishes, its results are marshalled and sent back to the client.

4.1 Probabilistic exploits

In the simulator security model, a vulnerability is a mechanism used to access an otherwise restricted communication channel. In this model, a real exploit payload is replaced by an ID or “magic string” which is sent to a simulated application. If this application is defined to be vulnerable (and some other requirements are fulfilled), then an agent will be installed in the computer hosting the vulnerable application.

The simulated exploit payload includes the aforementioned
magic string. When the Simulator subsystem receives this information, it looks up for the string in the Exploits Database. If it is found, then the simulator will decide if the exploit worked or not and with what effect based on a probability distribution that depends on the effective scenario information of that computer and the specific exploit. Suppose, for example, that the Penetration Testing Framework assumes (wrongly) the attacked machine is a Red Hat Linux 8.0, but that machine is indeed a Windows system. In this hypothetical situation, the exploit would fail with 100% of probability. On the other side, if the attacked machine is effectively running an affected version of Red Hat Linux 9.0, then the probability of success could be 75%, or as determined in the exploit database.

4.2 Remote attack model overview

In Fig. 2 we can see the sequence of events which occurs when an attacker launches a remote exploit against a simulated machine. The rectangles in the top are the four principal components involved: The Penetration Testing Framework, the Simulator and the Exploits Database are the subsystems explained in Fig. 1 the Vulnerable Application is a simulated application or service which is running inside an Insight scenario and has an open port. In the diagram the declared components are represented as named rectangles, messages are represented as solid-line arrows, and time is represented as a vertical progression.

When an exploit is launched against a service running in a simulated machine, a connection is established between the Penetration Testing Framework and the service. Then, the simulated exploit payload is sent to the application. The targeted application reads the payload by running the system call read. Every time the syscall read is invoked, the Simulator subsystem analyzes if a magic string is present in the data which has just been read. When a magic string is detected, the Simulator searches for it in the Exploits Database. If the exploit is found, a new agent is installed in the compromised machine.

The exploit payload also includes information of the OS that the Penetration Testing Framework knows about the attacked machine: OS version, system architecture, service packs, etcetera. All this information is used to compute the probabilistic function and allows the Simulator to decide whether the exploit should succeed or not.

4.3 Local attack model overview

Insight can also simulate local attacks: If an attacker gains control over a machine but does not have enough privileges to complete a specific action, a local attack can deploy a new agent with higher privileges.

5. DETAILED DESCRIPTION

One of the most challenging issues in the Insight architecture is to resolve the tension between realism and performance. The goal was to have a simulator on a single desktop computer, running hundreds of simulated machines, with a simulated traffic realistic from a penetration test point of view. But there is a trade-off between realism and performance and we are going to discuss some of these problems and other architecture details in the following sections.
5.1 The Insight development library

New applications can be developed for the simulation platform using a minimal C standard library, a standardized collection of header files and library routines used to implement common operations such as: input, output and string handling in the C programming language.

This library—a partial libc—implements the most common functions (e.g., read, write, open), allowing any developer to implement his own services with the usual compilers and development tools (e.g., gcc, g++, MS Visual Studio). For example, a web server could be implemented, linked with the provided libc and plugged within the Insight simulated scenarios.

The provided libc supports the most common system calls, but it is still incomplete and we were unable to compile complex open source applications. In spite of this, some services (e.g., a small DNS) and network tools (e.g., ipconfig, netstat) have been included in the simulation platform, and new system calls are planned to be supported in the future.

5.2 Simulating sockets

A hierarchy for file descriptors has been developed as shown in Fig. 4. File descriptors can refer (but they are not limited) to files, directories, sockets, or pipes. At the top of the hierarchy, the tree root shows the descriptor object which typically provides the operations for reading and writing data, closing and duplicating file descriptors, among other generic system calls.

![Descriptors’ object hierarchy tree.](image)

The simulated sockets implementation spans between two kinds of supported sockets subclasses:

1. SocketDirect. This variety of sockets is optimized for the simulation in one computer. Socket direct is fast: as soon as a connection is established, the client keeps a file descriptor pointing directly to the server’s descriptor. Routing is only executed during the connection and the protocol control blocks (PCBs) are created as expected, but they are only used during connection establishment. Reading and writing operations between direct sockets are carried out using shared memory. Since both sockets can access the shared memory area like regular working memory, this is a very fast way of communication.

2. SocketReal. In some particular cases, we are interested in having full socket functionality. For example, the communication between Insight and the outside world is made using real sockets. As a result, this socket subclass wraps a real BSD socket of the underlying OS.

Support for routing and state-less firewalling was also implemented, supporting the simulating of attack payloads that connect back to the attacker, accept connections from the attacker or reuse the attack connection.

5.3 The exploits database

When an exploit is raised, Insight has to decide whether the attack is successful or not depending on the environment conditions. For example, an exploit can require either a specific service pack installed in the target machine to be successful, or a specific library loaded in memory, or a particular open port, among others requirements. All these conditions vary over the time, and they are basically unpredictable from the attacker’s standpoint. As a result, the behavior of a given exploit has been modeled using a probabilistic approach.

In order to determine the resulting behavior of the attack, Insight uses the Exploits Database showed in the architecture layout of Fig. 1. It has a XML tree structure. For example, if an exploit succeeds against a clean XP professional SP2 with 83% probability, or crashes the machine with 0.05% probability in other case; this could be expressed as follows:

```xml
<database>
  <exploit id="sample exploit">
    <requirement type="system">
      <os arch="i386" name="windows" />
      <win>XP</win>
      <edition>professional</edition>
      <servicepack>2</servicepack>
    </requirement>
    <results>
      <agent chance="0.83" />
      <crash chance="0.05" what="os" />
      <reset chance="0.00" what="os" />
      <crash chance="0.00" what="application" />
      <reset chance="0.00" what="application" />
    </results>
  </exploit>
  ...
</database>
```

The conditions needed to install a new agent are described in the requirements section. It is possible to use several tags in this section, they specify the conditions which have influence on the execution of the exploit (e.g., OS required, a specific application running, an open port). The results
section is a list of the relevant probabilities. In order, these are the chance of:

1. successfully installing an agent,
2. crashing the target machine,
3. resetting the target machine,
4. crashing the target application,
5. and the chance of resetting the target application.

To determine the result, we follow this procedure: processing the lines in order, for each positive probability, choose a random value between 0 and 1. If the value is smaller than the chance attribute, the corresponding action is the result of the exploit.

In this example, we draw a random number to see if an agent is installed. If the value is smaller than 0.83, an agent is installed and the execution of the exploit is finished. Otherwise, we draw a second number to see if the OS crashes. If the value is smaller than 0.05, the OS crashes and the attacked machine becomes useless, otherwise there is no visible result. Other possible results could be: raising an IDS alarm, writing some log in a network device (e.g. firewall, IDS or router) or capturing a session id, cookie, credential or password.

The exploits database allows us to model the probabilistic behavior of any exploit from the attacker’s point of view, but how do we populate our database? A paranoid approach would be to assign a probability of success of 100% to every exploit. In that way, we would consider the case where an attacker can launch each exploit as many times as he wants, and will finally compromise the target machine with 100% probability (assuming the attack does not crash the system).

A more realistic approach is to use statistics from real networks. Currently we are using the framework presented by Marcelo Picorelli \[18\] in order to populate the probabilities in the exploits database. This framework was originally implemented to assess and improve the quality of real exploits in QA environments. It allows us to perform over 500 real exploitation tests daily on several running configurations, spanning different target operating systems with their own setups and applications that add up to more than 160 OS configurations. In this context, a given exploit is executed against:

- All the available platforms
- All the available applications

All these tests are executed automatically using low end hardware, VMware servers, OS images and snapshots. The testing framework has been designed to improve testing time and coverage, and we have modified it in order to collect statistical information of the exploitation test results.

5.4 Scheduler

The scheduler main task is to assign the CPU resources to the different simulated actors (e.g. simulated machines and process). The scheduling iterates over the hierarchy machine-process-thread as a tree (like a depth-first search), each machine running its processes in round-robin.

In a similar way, running a process is giving all its threads the order to run until a system call is needed. Obviously, depending on the state of each thread, they run, change state or finish execution. The central issue is that threads execute systems calls and then (if possible) continue their activity until they finish or another system call is required.

Insight threads are simulated within real threads of the underlying OS. Simulated machines and processes are all running within one or several working processes (running hundreds of threads), and all of them are coordinated by a unique scheduler process called the master process. Thanks to this architecture, there is a very low loss of performance due to context switching\[7\].

5.5 File system

In order to handle thousand of files without wasting huge disk space, the file system simulation is accomplished by mounting shared file repositories. We are going to refer these repositories as template file systems. For example, all simulated Windows XP systems could share a file repository with the default installation provided by Microsoft. These shared templates would have reading permission only. Thus, if a virtual machine needs to read or change a file, it will be copied within the local file system of the given machine.

This technique is well known as copy-on-write. The fundamental idea is allowing multiple callers asking for resources which are initially indistinguishable, giving them pointers to the same resource. This function can be maintained until a caller tries to modify its copy of the resource, at which point a true private copy is created to prevent the changes from becoming visible to everyone else. All of this happens transparently to the callers. The primary advantage is that no private copy needs to be created if a caller never makes any modification.

On the other hand, with the purpose of improving the simulator’s performance, a file cache has been implemented: the simulator saves the most recent accessed files (or block of files) in memory. In high scale simulated scenarios, it is very common to have several machines doing the same task at (almost) the same time\[8\]. If the data requested by these kind of tasks are in the file system cache, the whole system performance would improve, because less disk accesses will be required, even in scenarios of hundreds or thousands simulated machines.

6. PERFORMANCE ANALYSIS

To evaluate the performance of the simulator we run a test including a scenario with an increasing number of complete

---

\[7\] Because descriptors and pointers remain valid when switching from one machine to the other.

\[8\] For example, when the simulation starts up, all UNIX machines would read the boot script from /etc/initd file.
LANs with 250 computers each, simultaneously emulated. The tests only involve the execution of a network discovery on the complete LANs through a TCP connection to port 80. An original pen-testing module used for information was executed with no modifications, this was a design goal of the simulator, to use real unmodified attack modules when possible.

| Performance of the simulator | LANs | Computers | Time (secs) | Syscalls/sec |
|------------------------------|------|-----------|-------------|--------------|
| 1                            | 250  | 80        | 356         |              |
| 2                            | 500  | 173       | 236         |              |
| 3                            | 750  | 305       | 175         |              |
| 4                            | 1000 | 479       | 139         |              |

Table 2: Evolution of the system performance as the simulated scenario grows, running a network discovery module, connecting to a predefined port. This benchmark was run on a single Intel Pentium D 2.67Ghz, 1.43GB RAM.

We can observe the decrease of system calls processed per second as we increase the number of simulated computer as Insight was ran on a single real computer with limited resources. Nevertheless, the simulation is efficient because system calls are required on demand by the connections of the module gathering the information of the networks through TCP connections.

7. APPLICATIONS

We have created a playground to experiment with cyber-attack scenarios which has several applications. The most important are:

- **Data collection and visualization.** Having the complete network scenario in one computer allows an easy capture and log of system calls and network traffic. This information is useful for analyzing and debugging real pen-test tools and their behavior on complex scenarios. Some efforts have been made to visualize attack pivoting and network information gathering using the platform presented.

- **Pentest training.** Our simulation tool is already being used in Pentest courses. It provides reproducible scenarios, where students can practice the different steps of a pentest: information gathering, attack and penetrate, privilege escalation, local information gathering and pivoting.

  The simulation allows the student to grasp the essence of pivoting. Setting up a real laboratory where pivoting makes sense is an expensive task, whereas our tool requires only one computer per student (and in case of network / computer crash, the simulation environment can be easily reset). Configuring new scenarios, with more machines or more complex topologies, is easy as a scenario wizard is provided.

  In Pentest classes with Insight, the teacher can check the logs to see if students used the right tools with the correct parameters. He can test the students' ability to plan, see if they did not perform unnecessary actions. The teacher can also identify their weaknesses as pentesters and plan new exercises to work on these. The students can be evaluated: success, performance, stealth and quality of reports can be measured.

- **Worm Spreading Analysis.** The lightweight design of the platform allows the simulation of socket/network behavior of thousands of computers gives a good framework for research on worm infestation and spreading. It should be possible to develop very accurate applications to mimic worm behavior using the Insight C programming API. There are available abstract modeling [7] or high-fidelity discrete event [27] studies but no system call level recreation of attacks like we propose in this future application of the platform.

- **Attack Planning.** It can be used as a flexible environment to develop and test attack planning algorithms used in automated penetration testing based on attack graphs [12].

- **Analysis of countermeasures.** Duplication of the production configuration on a simulated staging environment accurately mimicking or mirroring the security aspects of an organization’s network allows the anticipation of software/hardware changes and their impact on security. For example, you can answer questions like “Will the network avoid attack vector A if firewall rule R is added to the complex rule set S of firewall F?”

- **Impact of 0-day vulnerabilities.** The simulator can be used to study the impact of 0-days (vulnerabilities that have not been publicly disclosed) in your network. How is that possible? We do not know current 0-days... but we can model the existence of 0-day vulnerabilities based on statistics. In our security model, the specific details of the vulnerability are not needed to study the impact on the network, just that it may exist with a measurable probability.

  That information can be gathered from public vulnerability databases: the discovery date, exploit date, disclosure date and patch date are found in several public databases of vulnerabilities and exploits [5, 23, 22, 11].

  The risk of a 0-day vulnerability is given by the probability of an attacker discovering and exploiting it. Although we do not have data about the security underground, the probabilities given by public information are a lower bound indicator.
As shown in [10], the risk posed by a vulnerability exists before the discovery date, augments as an exploit is made available for the vulnerability, and when the vulnerability is disclosed. The risk only diminishes as a patch becomes available and users apply the patches (and workarounds).

The probability of discovery, and the probability of an exploit being developed, can be estimated as a function of the time before disclosure (see Fig. 5 taken from [10]). For Microsoft products, we have visibility of upcoming disclosures of vulnerabilities: every month (on patch Tuesday) on average 9.40 patches are released (high and medium risk), based on those dates we estimate the probability that the vulnerabilities were discovered and exploited during the months before disclosure.

8. CONCLUSION
We have created a playground to experiment with cyber-attack scenarios. The framework is based on a probabilistic attack model—that model is also used by attack planning tools developed in our lab. By making use of the proxy syscalls technology, and simulating multiplatform agents, we were able to implement a simulation that is both realistic and lightweight, allowing the simulation of networks with thousands of hosts.

The framework provides a global view of the scenarios. It is centered on the attacker’s point of view, and designed to increase the size and complexity of simulated scenarios, while remaining realistic for the attacker.

The value of this framework is given by its multiple applications:

- Evaluate network security
- Evaluate security countermeasures
- Anticipate the risk posed by 0-day vulnerabilities
- Pen test training
- Worm spreading analysis
- Systematic study of Planning techniques
- Data generation to test visualization techniques

If you are interested in using Insight, send us an email. We are trying to build a community using it as common language for discussing information security scenarios and practices, and will strongly support new applications of this tool.

9. REFERENCES
[1] Chris Anley, John Heasman, Felix Lindner, and Gerardo Richarte. The Shellcoder’s Handbook. Wiley Press, 2nd edition, 2007.
[2] Michael Bailey, Evan Cooke, Farnam Jahanian, Jose Nazario, and David Watson. The internet motion sensor: A distributed blackhole monitoring system. In In Proceedings of Network and Distributed System Security Symposium NDSS ’05, pages 167–179, 2005.
[3] S. M. Bellovin. Security problems in the TCP/IP protocol suite. Computer Communications Review, 19:32–48, 1989.
[4] Rainer Bye, Stephan Schmidt, Katja Luther, and Sahin Albayrak. Application-level simulation for network security. In Proceedings of the First International Conference on Simulation Tools and Techniques for Communications, Networks and Systems, 2008.
[5] Maximiliano Caceres. Syscall proxying - simulating remote execution. Technical report, CoreLabs, Core Security Technology, 2002. Available from http://www.coresecurity.com.
[6] CERT. Computer Emergency Response Team, USA. http://www.cert.org.
[7] Z. Chen, L. Gao, and K. Kwiat. Modeling the spread of active worms. In Proceedings of IEEE INFOCOM 2003, 2003.
[8] R. Davoli. VDE: virtual distributed Ethernet. In First International Conference on Testbeds and Research Infrastructures for the Development of Networks and Communities, volume 1, pages 213–220, 2005.
[9] Jeff Dike. User Mode Linux. Prentice Hall, 1st edition, 2006.
[10] Stefan Frei, Martin May, Ulrich Fiesler, and Bernhard Plattner. Large-scale vulnerability analysis. In LSAD ’06: Proceedings of the 2006 SIGCOMM workshop on Large-scale attack defense, pages 131–138, New York, NY, USA, 2006. ACM.
[11] FrSirt. French Security Incident Response Team, France. http://www.fr-sirt.com.
[12] Ariel Futoransky, Luciano Notarfrancesco, Gerardo Richarte, and Carlos Sarraute. Building computer network attacks. Technical report, CoreLabs, Core Security Technology, 2003. Available from http://www.coresecurity.com.
[13] Michael Liljenstam and Jason Liu and David Nicol and Anthony Shuff and Stuart Staniford and Nicholas Weaver. Rinse: The real-time immersive network simulation environment for network security exercises. In Workshop on Principles of Advanced and Distributed Simulation, 2005.
[14] Jean-Vincent Loddo and Luca Saiu. Marionnet: A virtual network laboratory and simulation tool. In First International Conference on Simulation Tools and Techniques for Communications, Networks and Systems, 2008.
[15] David Moore, Vern Paxson, Stefan Savage, Colleen Shannon, Stuart Stanford, and Nicholas Weaver. Inside the slammer worm. IEEE Security and Privacy, 1(4):33–39, 2003.
[16] H. D. Moore. Metasploitation. In CanSecWest 2006, 2006.
[17] Aleph One. Smashing the stack for fun and profit. Phrack, 49–4, nov 1996. Available from http://www.phrack.com.
[18] Marcelo Picorelli. Virtualization in software development and QA, 2006. WMWORLD 2006 - http://www.wmwORLD.com.
[19] The Honeynet Project. Know your enemy: Learning about security threats. Addison-Wesley Professional,
2nd edition, 2004.

[20] The Honeynet Project. Know your enemy: honeynets. Technical report, Infocus At Securityfocus.com, May 2006. [http://www.honeynet.org/papers/honeynet/](http://www.honeynet.org/papers/honeynet/)

[21] Niels Provos. A virtual honeypot framework. In *Proceedings of the 13th USENIX Security Symposium*, pages 1–14, 2004.

[22] Secunia. [http://www.secunia.com](http://www.secunia.com)

[23] SecurityFocus. [http://www.securityfocus.com](http://www.securityfocus.com)

[24] D. Song, R. Malan, and R. Stone. A snapshot of global internet worm activity. Technical report, Arbor Networks, Nov 2001.

[25] L. Spitzner. *Honeypots: Tracking Hackers*. Addison-Wesley Longman Publishing Co., Inc., Boston, MA, USA, 2002.

[26] Michael Vrable, Justin Ma, Jay Chen, David Moore, Erik Vandekieft, Alex C. Snoeren, Geoffrey M. Voelker, and Stefan Savage. Scalability, fidelity, and containment in the potemkin virtual honeyfarm. *SIGOPS Oper. Syst. Rev.*, 39(5):148–162, 2005.

[27] Songjie Wei, Jelena Mirkovic, and Martin Swany. Distributed worm simulation with a realistic internet model. In *PADS ’05: Proceedings of the 19th Workshop on Principles of Advanced and Distributed Simulation*, pages 71–79, Washington, DC, USA, 2005. IEEE Computer Society.

[28] V. Yegneswaran, P. Barford, and D. Plonka. The design and use of internet sinks for network abuse monitoring. In *Proceedings of Recent Advances in Intrusion Detection (RAID)*, Sept. 2004.