Identification of potentially harmful requests directed at web sites

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The sad state of web security

- IVIZ: 99% of tested applications (5000) have at least one vulnerability, 35 on average
- WhiteHat: 86% of tested applications are vulnerable
- Symantec: 78% websites with vulnerabilities, 16% critical
 - In just one month (05.2012) LizaMoon was responsible for at least a million successful SQL Injection attacks





Scope of the method

- There are several classes of the attacks
- The method discussed focuses on the abuse of path traversal:
 - Multiple attempts
 - Forceful browsing
 - Unusual usage patterns



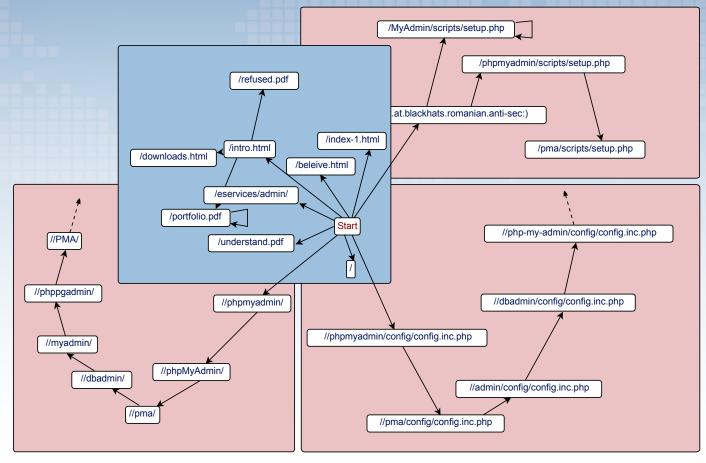


Part 1 Behavior modeling





A real-life scenario







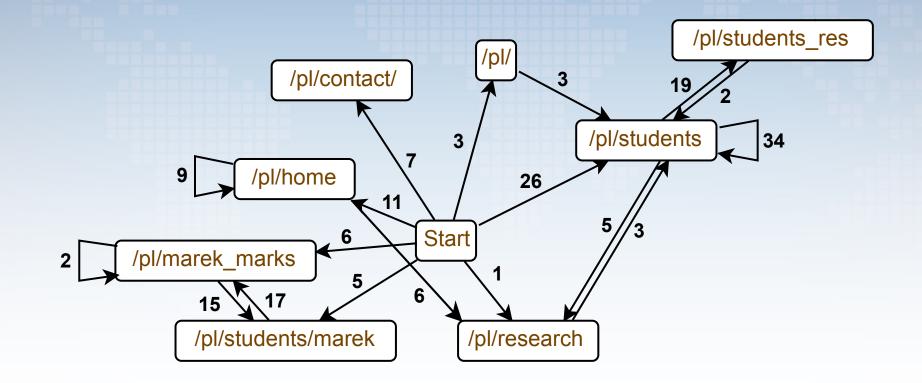
Behavior analysis

- With large enough sample of data it is possible to identify typical usage patterns
- These patterns can be derived from the frequency of transitions between pages (URLs), which in turn can be represented as weights of a graph's edges
- Unusual behavior can be detected by the lack of supporting graph edge (or its low weight)





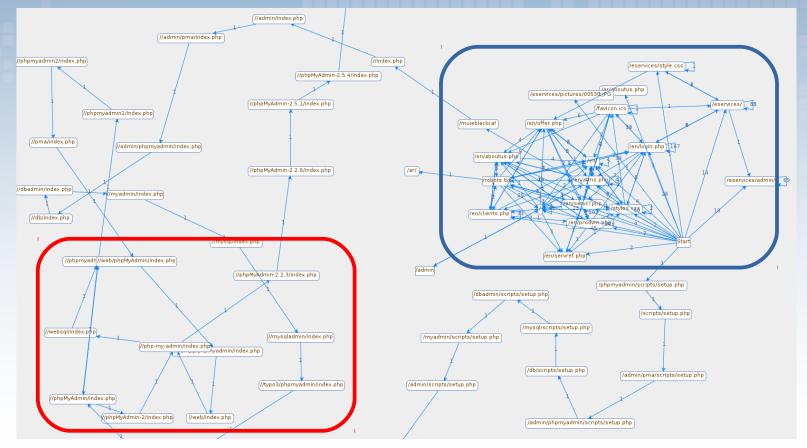
Sample log-based graph







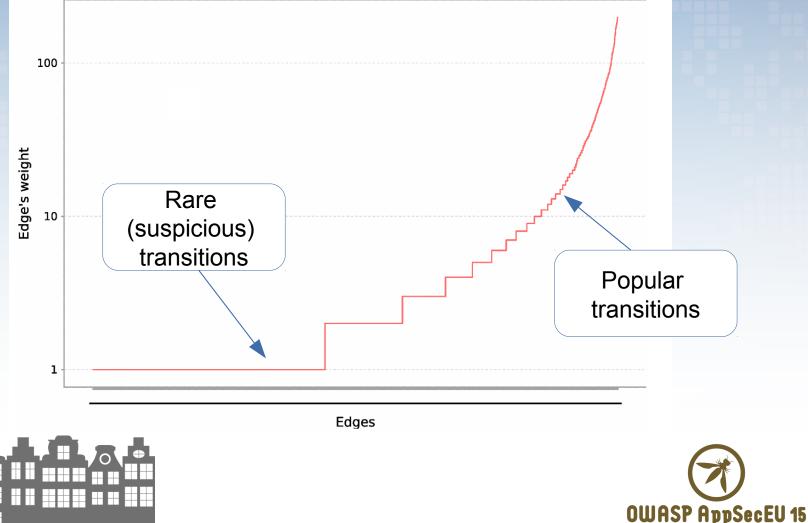
Another sample transition graph







Distribution of edges' weights



UWHSP HPPSECE Amsterdam, The Netherlands

Benefits of the graph structure

- Effective representations of large log data sets
- On-line analysis and detection of (some) suspicious behavior
- Identification of unusual requests, marking of potentially harmful requests
- Also, optimization of site's performance:
 - Pre-caching of expected web pages
 - Path-to-target minimization





Problems & issues

- Assembly of sessions from the requests
 - SID is the most reliable, but not always available
- Filtering out of follow-up requests (CSS, images)
- 'Unusual' but legitimate transitions
 - Introduction of 'trust earned'





Part 2 Collective assessment of suspicious requests





Rationale

- Lot of attack attempts are more or less automated. They are performed by malware or 'script-kiddies'
- As a result, (near) identical requests appear at different servers
- If the servers share the information about such requests, they can identify attack attempts with greater accuracy





Securing the confidentiality

- The requests (URLs) cannot be made publicly available, as this could lead to a leak of sensitive information
- However, it is sufficient to publish and share a cryptographic hash (e.g. MD5) of the request URL
- URL needs to be stripped of server specific part (domain, etc.), and likely request parameters





Sample data exchange format

{	C:O,	T:M,	A:57,
{	C:O,	T:M,	A:53,
{	C:F,	T:M,	A:32,
{	C:F,	T:M,	A:31,
{	C:F,	Τ:Μ,	A:24,
{	C:O,	Τ:Β,	A:17,

MD5:2cf1d3c7fe2eadb66fb2ba6ad5864326 }
MD5:2370f28edae0afcd8d3b8ce1d671a8ac }
MD5:2f42d9e09e724f40cdf28094d7beae0a }
MD5:8f86175acde590bf811541173125de71 }
MD5:eee5cd6e33d7d3deaf52cadeb590e642 }
MD5:bd9cdbfedca98427c80a41766f5a3783 }

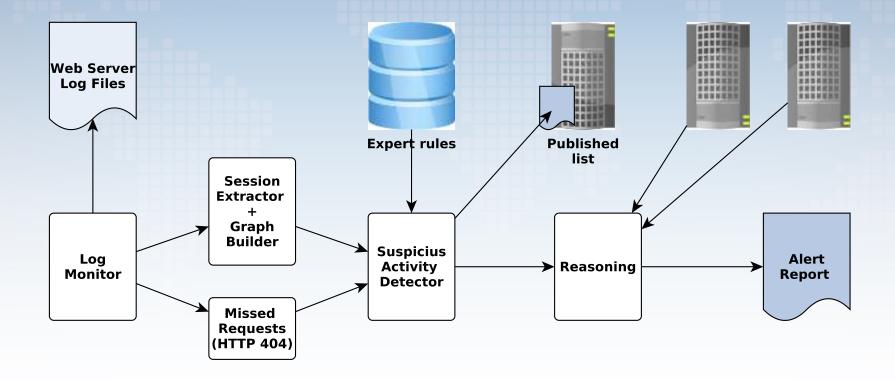
Additional algorithm-tuning information

Age of the event

<u>Type of report</u> **M**: missing resource **B**: Behavioral anomaly



Reference implementation







Initial results

- Tests ran for a year on three (small) servers
- Approximately 30% of the attacks detected (compared to semi-manual log analysis)
- The ratio is expected to raise with the number of servers involved
- The results required no a-priori training
- Most importantly: no false positives





Thank you for your attention

Any questions are welcome



