Modeling and Containment of Search Worms Targeting Web Applications

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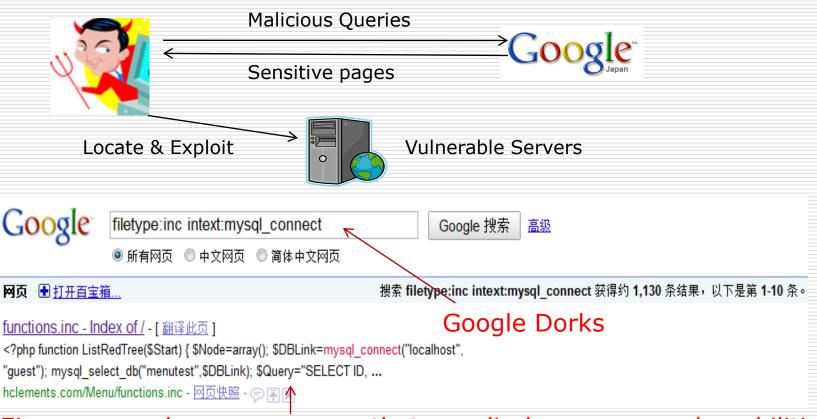
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Outline

- □ Background: What are Search Worms?
- Motivation: Modeling & Containing
- Modeling of Search Worms
- Containment of Search Worms
- Conclusion
- ☐ Future Work

Background-Google Hacking

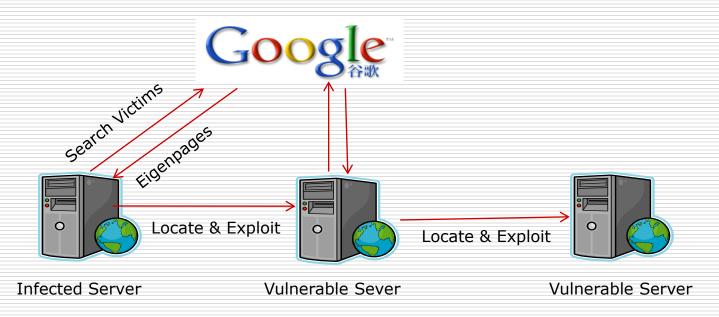
Google is a great tool for Hackers!!!



Eigenpages: dangerous pages that can disclose server vulnerabilities

□ 2010/7/9 DIMVA'10

Background-Search Worm



Search Worm: a worm uses search engines to locate targets

Example:

Santy: targeting phpBB bulletin system.

Google Dork: allinurl: viewtopic.php

Released on 12/20/2004, infected 40,000 servers in 2 days

Motivation

- Search Worms are dangerous! So, It deservers our research.
- Two common problems in the field of worm studying:
 - How to model the propagation of such worms ? (I(t) = ?)
 - Study the spreading characteristics
 - ☐ Study the effects of containment strategies
 - How to contain the propagation of such worms?

Topic 1 Modeling of Search Worms

Modeling of the Search Worms A virtual search worm

Vulnerable severs leak eigenpages containing specific keywords to search engine.

N: num of severs containing the eigenpages (Suspicious severs)

V: num of really exploitable severs (Vulnerable servers)

- Propagation steps of infected severs:
 - 1. Search "special-keywords and random-keywords" in a search engine $\rightarrow m$ search results
 - 2. Choose δ pages among the total m search results to scan.
 - 3. Once a server is infected, it begins this infection cycle, too.

- Obviously, severs contain more eigenpages are more likely to be exploited.
- Two attacker-favorable assumptions:
 - m search results are randomly selected from all the eigenpages on the web
 - lacksquare δ targets are randomly selected from the m search results

U-Model: eigenpages are uniformly distributed on severs

The number of infected servers by the end of time tick *t*:

$$I(t) = I(t-1) + \underbrace{[V-I(t-1)]}_{\text{Remained vulnerable severs by the end of the time tick t}^{\text{Newly infected}} \xrightarrow{\text{severs during the time tick t}}^{\text{Newly infected}} \xrightarrow{\text{severs during the time tick t}}^{\text{Newly infected}} \xrightarrow{\text{severs during the time tick t}}^{\text{Newly infected}}$$

V is the total count of servers really suffering the vulnerabilities among the N severs containing eigenpages

PL-Model: eigenpages follow a power law distribution.

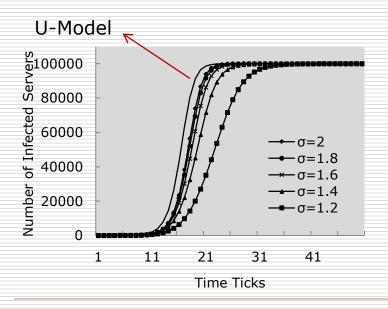
The probability that the number of eigenpages p on a suspicious server is greater than x is

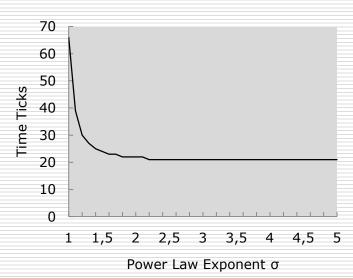
$$a_i = N \cdot [prob(p > p_i) - prob(p > p_{i+1})] = N \cdot [(\frac{p_{min}}{p_i})^{\sigma} - (\frac{p_{min}}{p_{i+1}})^{\sigma}]$$

PL-Model:

Newly infected severs in the *i*-th group during the time tick t

$$\begin{cases} I(t,i) = I(t-1,i) + [V_i - I(t-1,i)][1 - (1 - \frac{p_i}{P})^{\delta I(t-1)}] \\ I(t) = \sum_{i=1}^k I(t,i) \end{cases}$$





Time to infect 95% vulnerable severs for the PL-Model

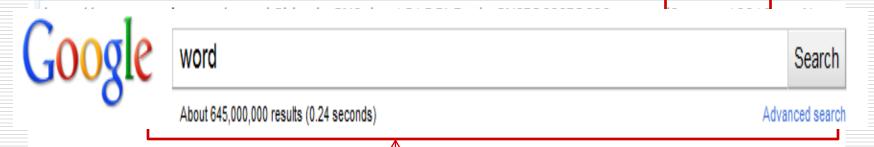
Proposition:

Among different distributions of eigenpages, the uniform distribution optimizes the performance of search worms.

We proved this conclusion by using the mean value inequality

Modeling of the Search Worms Effects of Page Ranking

- Search results are ranked according to Keyword relevance, page importance....
- □ Pages on popular servers are more likely to appear in front→ scan collisions.
- If the second attacker-favorable assumption is true, no scan collisions.



However, the second assumption is impossible, the propagation of search worm will be affected by page ranking

Modeling of the Search Worms Effects of Page Ranking

- □ Page importance:
 - Page Ranking Value (0-10)
 Power Law [Litvak 2007]

$$prob(PR=k) = \begin{cases} (\frac{1}{6^k})^\alpha - (\frac{1}{6^{k+1}})^\alpha & 0 \leq k < 10 \\ (\frac{1}{6^{10}})^\alpha & k = 10 \end{cases}$$

Site importance
 We simply assume servers contain more pages are more important

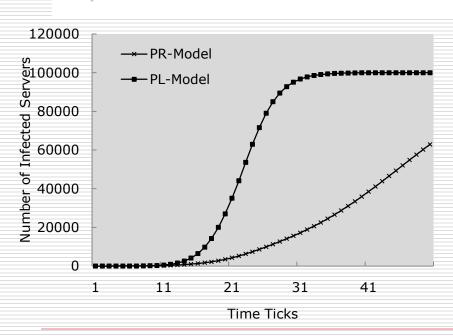
[Litvak 2007] Litvak, N., Scheinhardt, W. R. W., Volkovich, Y.: In-degree and PageRank: Why do they follow similar power laws? Internet Math, Vol.4(2-3), pp.175-198 (2007)

Modeling of the Search Worms Effects of Page Ranking

PR-Model:

An infected server selects the δ top-ranking

$$\begin{cases} I(t,i) = I(t-1,i) + [V_i - I(t-1,i)][1 - (1 - \frac{1}{a_i})^{\delta_i I(t-1)}] \\ I(t) = \sum_{i=1}^k I(t,i) \end{cases}$$



Page ranking slows down the spreading of the search worm

Topic 2: Containment of Search Worms

- The goal to model the search worm is to help developing an efficient containment system.
- We introduce a conceptual containment system based on honey-page insertion.
- We use our propagation models to analyze this system.

☐ Honey Page

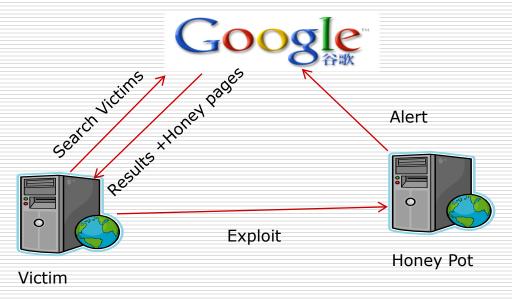


This is a honey page pointing to a honey pot

Containment based on honey-page insertion:

Search engine randomly inserts honey-pages into search results for

any query.



When the search engine receives an alert, it denies further queries form detected victims

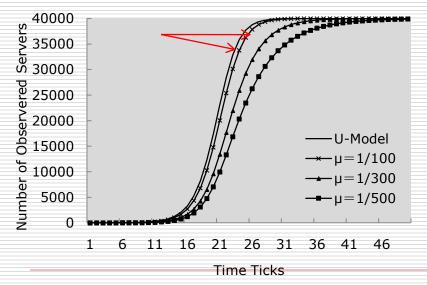
Is such a strategy possible?

↑ Honey page insert rate

$$D_{t+1} = D_t + (I_t - D_t) [1 - (1 - \mu)^{\delta}]$$

of detected infected nodes by the time t+1

Probability that an infected sever scans a honey page during an infection cycle



An infected sever can be induced to scan honeypots in a very short time after it is infected even if the insert rate is very small.

- Two questions:
 - Containment requirement → Insert Rate?

$$\mu=1-[1+\frac{\gamma\varepsilon\delta}{\ln{(1-\gamma)}}]^{1/\delta} \quad \mbox{γ is the final prevalence rate:} \\ \mbox{ration of infected vulnerable} \\ \mbox{servers}$$

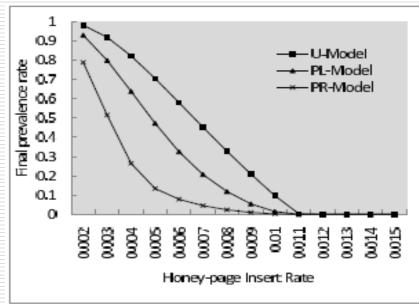
Is arbitrary requirement can be reached?

 $\varepsilon\delta \leq 1$ No limitation, arbitrary requirement can be reached $\varepsilon = \frac{V}{N} \text{ is the density of vulnerable servers.}$

Effectiveness for the Santy worm

To contain the final prevalence rate below 1%

Insert rate \rightarrow 0.011



2 honey pages in every 100 search results can stop the spreading of the Santy worm at its early age!

Conclusion

- Modeling of the Search Worm
 - Eigenpages Distribution: Uniform distribution optimizes the spreading
 - Page Ranking: slow downs the spreading
- Containment of the Search Worm
 - Honey page insertion
 - A small insert rate can lead a good containment effect

Challenging Future Work

Worm may validate the truth of the search results. Then, how to disguise honey pages as true ones both in URL and contents?

Our current conclusions are based on simulation. Real experiments are required to verify them.

Question & Answer

Thanks for your attention!