## Security Analyses



## One-Slide Summary

- We can statically detect buffer overruns in programs by modeling the space allocated for a buffer and the space used for a buffer. We cannot be right all the time.
- SQL injection and cross-site scripting attacks occur when evil user input is used (parsed) as part of another important language (e.g., HTML or SQL).
- Program analyses are expensive; recent research can randomize them to save time.


## Lecture Outline

- Static Analyses to Detect Buffer Overruns
- Strings
- Alloc, Used
- Constraints
- SQL Injection Attacks
- Untrusted User Strings
- Interpreted as valid SQL
- Randomized Dataflow Analysis
- Random Join

> Static Analysis to Detect Buffer Overruns

- Detecting buffer overruns before distributing code would be better
- Idea: Build a tool similar to a type checker to detect buffer overruns
- This is a popular research area; we'll present one idea at random [Wagner, Aiken, ...]
- You'll see more in later lectures


## Focus on Strings

- Most important buffer overrun exploits are through string buffers
- Reading an untrusted string from the network, keyboard, etc.
- Focus the tool only on arrays of characters



## Idea 1: Strings as an Abstract Data Type

- A problem: Pointer operations and array dereferences are very difficult to analyze statically
- Where does *ptr point?
- What does buf[j] refer to?
- Idea: Model effect of string library functions directly
- Hard code effect of strcpy, strcat, etc.


## Idea 2: The Abstraction

- Model buffers as pairs of integer ranges
- Alloc min allocated size of the buffer in bytes
- Used max number of bytes actually in use
- Use integer ranges
- $[x, y]=\{x, x+1, \ldots, y-1, y\}$
- Alloc and used cannot be computed exactly


## The Strategy

- For each program expression, write constraints capturing the alloc and used of its string subexpressions
- Solve the constraints for the entire program
- Check for each string variable s

$$
\text { used(s) } \leq \operatorname{alloc}(\mathrm{s})
$$

## The Constraints

char s[n];
$\mathrm{n}=$ alloc(s)
strcpy(dst,src)
$p=s t r d u p(s)$
used(s) $\leq$ used(p) \&
alloc(s) $\leq$ alloc(p)
$\mathrm{p}[\mathrm{n}]=$ ' $\backslash 0^{\prime}$
$\min ($ used $(\mathrm{p}), \mathrm{n}+1)) \leq$
used(p)

## Constraint Solving

- Solving the constraints is akin to solving dataflow equations
- Remember liveness? Constant prop?
- Build a graph
- Nodes are len(s), alloc(s)
- Edges are constraints len(s) $\leq \operatorname{len}(\mathrm{t})$
- Propagate information forward through the graph
- Special handling of loops in the graph


## Results

- This technique found new buffer overruns in sendmail
- Which is like shooting fish in a barrel ...
- Found new exploitable overruns in Linux nettools package
- Both widely used
- Previously hand-audited packages



## Limitations

- Tool produces many false positives (why?)
- 1 out of 10 warnings is a real bug
- Tool has false negatives (why?)
- Unsound: may miss some overruns
- But still productive to use
- So let's pretend we used it ...


## Cat and Mouse

- Suppose I have a server (e.g., Amazon.com)
- Let's imagine that I have solved ...
- Viruses: no malicious code on machine
- Buffer overruns: no injection of evil assembly code
- Buffer overruns: no non-control data attacks
- Privileges: no running at root
- Spam: as long as I'm dreaming, l'd like a pony ...
- I can still convince the server to do the wrong thing with the resources it legitimately has access to ...


## Three-Tier Web Application

- This is how Amazon is structured
- Query is a SQL database command generated by program logic


## Presentation tier

The top-most level of the application is the user interface. The main function of the interface is to translate tasks and results to something the user can understand.

## Logic tier

This layer coordinates the application, processes commands, makes logical decisions and evaluations, and performs calculations. It also moves and processes data between the two surrounding layers.

## Data tier

Here information is stored and retrieved from a database or file system. The information is then passed back to the logic tier for processing, and then eventually back to the user.


## The Problem In The Logic Tier

```
$userid = read_from_network();
if (!eregi('[0-9]+', $userid)) {
    unp_msg('You entered an invalid user ID.');
    exit;
}
$user = $DB->query("SELECT * FROM `unp_user`". "WHERE userid='\$userid'");
if (!DB->is_single_row($user)) {
        unp_msg('You entered an invalid user ID.');
        exit;
```


## The Problem



## The Bad Place

```
\Gamma// $userid == "1'; DROP TABLE unp_user; --"
if (!eregi('[0-9]+', $userid)) {
    unp_msg('You entered an invalid user ID.');
    exit;
}
\$user = \$DB->query("SELECT * FROM `unp_user`". "WHERE userid='\$userid'");
if (!DB->is_single_row(\$user)) \{ unp_msg('You entered an invalid user ID.'); exit;
```


## The Bad Place: Destroying Data



## Also A Bad Place: Viewing Data

## SQL Code-Injection Vulnerabilities

- A SQL injection attack exploits a vulnerability in the database layer of an application whereby user input is incorrectly filtered for string literal escape characters or otherwise unexpected executed.
- Most common types of vulnerability in 2006:
- 25.1\% Cross-Site Scripting
- 14\% SQL Command Injection
- 7.9\% Buffer Overruns
- Attacks are easy and expose valuable data


## Exploits Of A Mom

- The essence of SQL injection:

HI, THIS IS YOUR SON'S SCHOOL. WE'RE HAVING SOME COMPUTER TROUBLE.



WELL, WEVE LOST THIS YEAR'S STUDENT RECORDS. I HOPE YOU'RE HAPPY.


AND I HOPE
YOUVE LEARNED
TO SANITIZE YOUR
DATABASE INPUTS.

## SQL Injection

- Note that it's basically a parsing problem
- We have a string constant in PHP plus a string constant from the user, and when combined they must make a valid SQL program
- One Solution: Dynamic Taint Analysis
- Propagate a "taint" bit with every string
- One Solution: Dynamic Grammar Analysis
- Partially parse PHP string fragment
- If PHP string fragment + user string fragment parses to something with a different top-level structure, bail!


## Parse Trees To The Rescue!

- Do the user input strings contribute to something "too high" on the parse tree?

(a)

(b)


## Cross-Site Scripting

- Cross-Site Scripting (XSS) has the same flavor
- Evil User X posts a message with JavaScript in it (e.g., send passwords to me) to Blog B
- Blog B can also be a forum, etc.
- Later, Luser browses Blog B
- Blog B sends over page data, including Evil X's Message
- Luser thinks it is from Blog B (misplaced trust)
- Luser renders and interprets it


## Stopping Evil Posts

- Evil network-crawling robots try to post evil JavaScript to every forum they can find
- Let's require a real human when posting
- Increases cost
- CAPTCHA
- Complete Automated
- Public Iuring test
- to tell Computers
- and Humans Apart


Result from image

## Have We Won Yet?

- CAPTCHAs fail in theory and in practice
- The overarching problem is exactly the same:
- The server takes input from an untrusted user
- That input may be interpreted by another parser later
- In SQL-CIVs, by the database's SQL parser
- In XSS, by a user's JavaScript parser
- So all of the same techniques apply for XSS


## Random Interpretation Sumit Gulwani \& George Necula



## Probabilistically Sound Program Analysis!

- Sound program analysis is hard (Rice's Theorem)
- PL researchers usually pay in terms of
- Loss of completeness or precision
- Complicated algorithms
- Long running times
- Can we pay in terms of soundness instead?
- Basically, soundness = correctness
- Judgments are unsound with low probability
- We can predict and control the probability of error
- Can gain simplicity and efficiency


## Discovering Affine Equalities

- Given a program (control-flow graph) ...
- Discover equalities of the form $2 y+3 z=7$
- Compiler Optimizations
- Loop Invariants
- Translation Validation
- There exist polynomial time deterministic algorithms [Karr 76]
- involving expensive operations - $\mathrm{O}\left(\mathrm{n}^{4}\right)$
- We present a randomized algorithm
- as complete as the deterministic algorithms
- but faster - O( $\mathrm{n}^{2}$ )
- and simpler (almost as simple as an interpreter)


## Example 1



## Example 1



- Random testing will have to exercise all the 4 paths to verify the assertions
- Our algorithm is similar to random testing
- However, we execute the program once, in a way that it captures the "effect" of all the paths
assert $(c+d=0)$; assert $(e=a+1)$


## Example 1



- Random testing will have to exercise all the 4 paths to verify the assertions
- Our algorithm is similar to random testing
- However, we execute the program once, in a way that it captures the "effect" of all the paths
- Exponential work, linear time! ( $P=N P$ ?)
$\operatorname{assert}(c+d=0)$; assert $(e=a+1)$


## Idea \#1: Affine Join Operation

- Execute both the branches
- Combine the values of the variables at joins using the affine join operation $\oplus_{w}$ for some randomly chosen w

$$
\mathrm{v}_{1} \oplus_{\mathrm{w}} \mathrm{v}_{2}=\mathrm{w} \times \mathrm{v}_{1}+(1-\mathrm{w}) \times \mathrm{v}_{2}
$$



## Example 1



- Choose a random weight for each join independently.
- All choices of random weights verify the first assertion
- Almost all choices contradict the second assertion.


## Example 1



- Choose a random weight for each join independently.
- All choices of random weights verify the first assertion
- Almost all choices contradict the second assertion.


## Example 1


$\operatorname{assert}(c+d=0) ; \operatorname{assert}(c=a+1)$

## Example 1



## Geometric Interpretation of the Affine Join operation

- : State before the join
- : State after the join

- satisfies all the affine relationships that are satisfied by both (e.g. $x+y=1, z=0$ )

Given any relationship that is not satisfied by any of 0 (e.g. $x=2$ ), also does not satisfy it with high probability

## Example 2



## Idea \#2: Adjust Operation

- Execute multiple runs of the program in parallel
- "Sample" = Collection of states at each program point
- "Adjust" the sample before a conditional (by taking affine joins of the states in the sample) such that
- Adjustment preserves original relationships
- Adjustment satisfies the equality in the conditional
- Use adjusted sample on the true branch


## Geometric Interpretation of the

 Adjust Operation

## The Randomized Interpreter R



## Completeness and soundness of $R$

- We compare the randomized interpreter R with a suitable actual interpreter A
- Actual Interpreter A would be too slow (etc.) to use in real life!
- R mimics A with high probability
- R is as complete as A
- R is sound with high probability


## Soundness Theorem

- If $A \Rightarrow g=0$, then with high probability $R \notin g=0$
- Error probability $\leq(2 d)^{b}\left(\frac{j+1}{d}\right)^{r}$
- b: number of branches
- j: number of joins
- d: size of the field
- r: number of points in the sample
- If $j=b=10, r=15, d \approx 2^{32}$, then error probability $\leq \frac{1}{2^{98}}$


## Conclusions, Wessy Summary

- Randomization can help achieve simplicity and efficiency at the expense of making soundness probabilistic
- Has been extended to handle uninterpreted function symbols, interprocedural analyses, randomized decision procedures for theorem proving, combined abstract interpreters, ...
- May help with complicated security analyses
- Go to grad school!


## Homework

- Final Exam Soon ...


