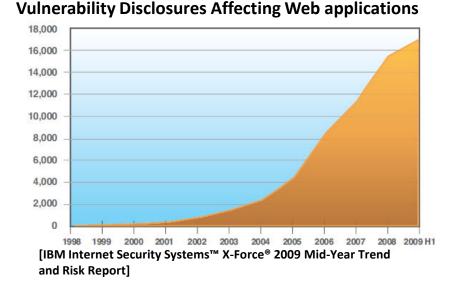
Security Testing of Web Applications: a Search Based Approach for XSS Vulnerabilities

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Background



One of the most prominent vuln class is: Cross-Site Scripting (XSS)

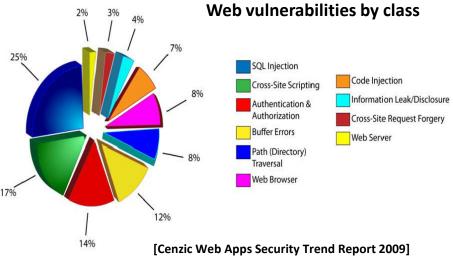
XSS causes the attacker to inject malicious code into the victim's browser

These vulnerabilities are due to missing or inadequate user input validation



Web applications are critical in many activities

Security of Web apps is also critical Number of total vulnerabilities in web applications is getting higher year by year

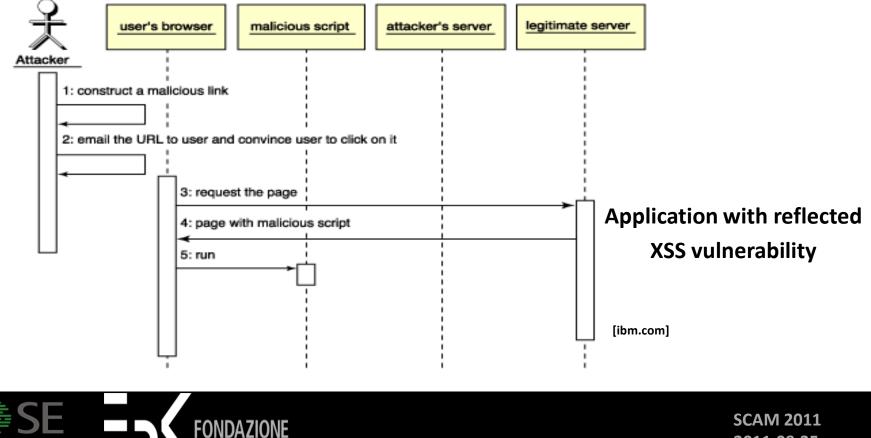


Background (II)

Attacker sets the trap – craft malicious link

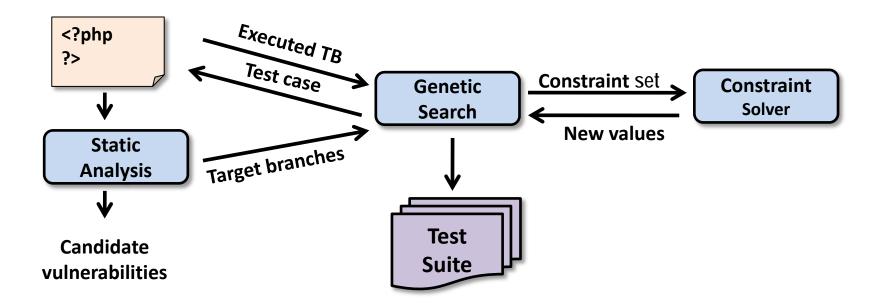
SOFTWARE

<A HREF=http://legitimateSite.com/registration.cgi?clientprofile=<SCRIPT>malicious code</SCRIPT>>Click here



2011.09.25

Our Proposal

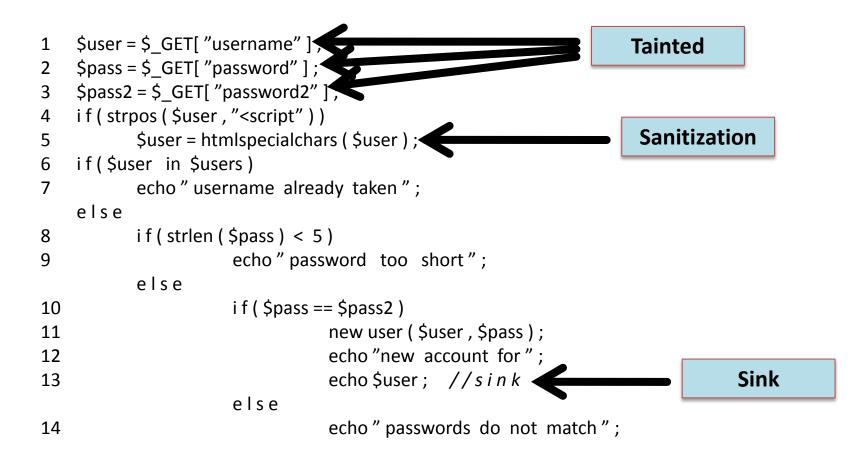


Three macro-blocks:

- **Static analysis** (vulnerabilities, target branches)
- Genetic algorithm (Global search: test case generation)
- Constraint solver (Local search: test case refinement)

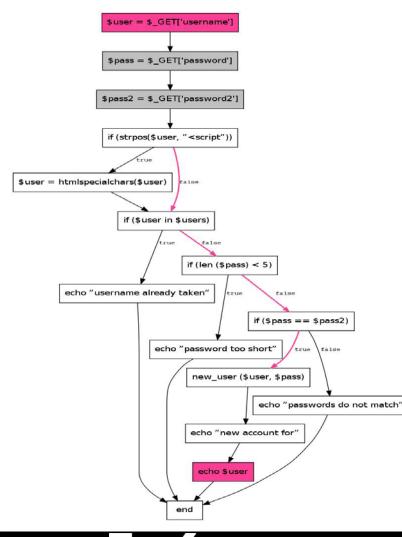


Running Example



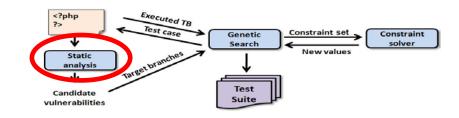


Static Analysis



FONDAT

ENGINEERING



Taint Analysis returns {\$user@1, \$user@13} as assignment chain

This means that

- there exists an input vector that flows to a sink without being sanitized (i.e. there is a vulnerability)
- Statements 1, 13 must be executed to trigger the vulnerability

On top of this information, we calculate control dependencies of the statements in the AC

{4-6, 6-8, 8-10, 10-11}

as target branches to execute

Genetic Algorithm



2 for (T in vulnerabilities) {

7

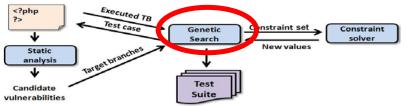
3 while (not covered (T) AND attempt < maxTry) {

```
4 selection = select ( population ) ;
```

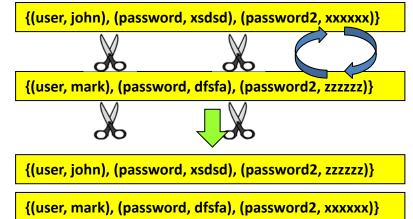
5 offspring = crossOver (selection) ;

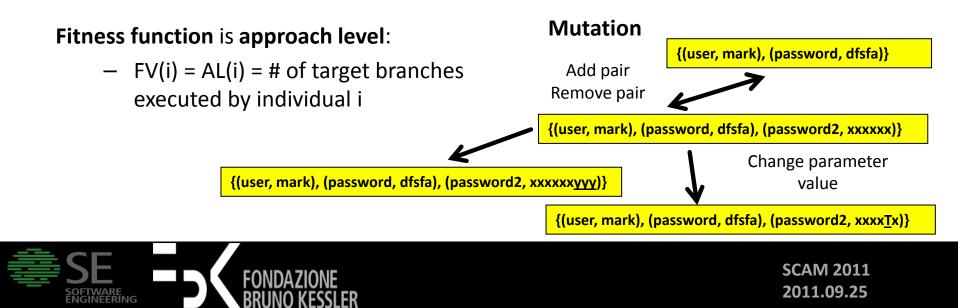
```
6 population = mutate ( offspring ) ;
```

```
attempt = attempt + 1 ;
```

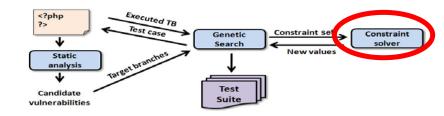


Crossover





Constraint Solving



When GA is not able to find a solution, a **constraint solver** is resorted

i = {(user, "ddeerer"), (password, "xxsdsed"), (password2, "dded33e")}

Branch	Condition	Target Branch
4-6	! strpos(GETusername, "< script")	4-6
6-8	false	6-8
8-10	! strlen(GETuser) < 5	8-10
10-14	! GETpassword == GETpassword2	10-11

Diverging point is calculated (branch 10-14) and respective constraint is negated ! strpos(GETusername, "<script") AND ! strlen(GETusername) < 5 AND GETpassword == GET password2 is passed to solver which could generate:

i_1 = {(user, "ddeerer"), (password, "dsfnggg"), (password2, "dsfnggg")}



Empirical Results

Run the tool on a real world application

Case study: Yapig 0.95-b

- Open source image gallery app, 53 files, 9 kloc

Page	#	Target Branches Covered			
		0%	38-46%	50-75%	100%
add_comment	1			1	
add_gallery	6		4	1	1
admin	1	1			
delete_gallery	4	2		1	1
modify_gallery	6	3		1	2
modify_phid	6	3		1	2
Slideshow	9				9
Upload	3		2		1
View	2	1		1	
Total	38	10	6	6	16

Cardinality	#	Target Branches Covered			
		0%	38-46%	50-75%	100%
1-2	15	7		3	5
3	7	2			5
4	7	1		2	4
5-13	9		6	1	2
Total	38	10	6	6	16



Advantages

- Static analysis is over-conservative
 No false negatives
- Search space is usually very large but GA heuristic helps in reducing it (global search)
- With the reduced search space, resorting to a constraint solver does not create scalability issues (local search)
- Actual executable test cases are generated for web applications



Limitations

- Static analysis is over-conservative
 False positives
- GA does not always converge to a solution
- Constraint solving is limited by the use of concrete values when:
 - Symbolic value is not always available or
 - Expressiveness of solver is limited
- Generated test cases are not actual attacks, they do not try to inject malicious code in the final page



Thanks for your attention!

