

Capture of silent security patches and reports

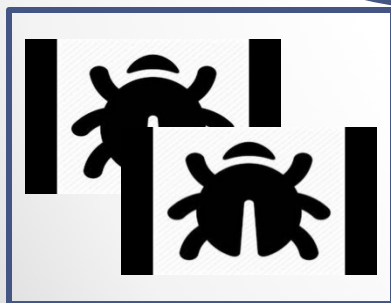
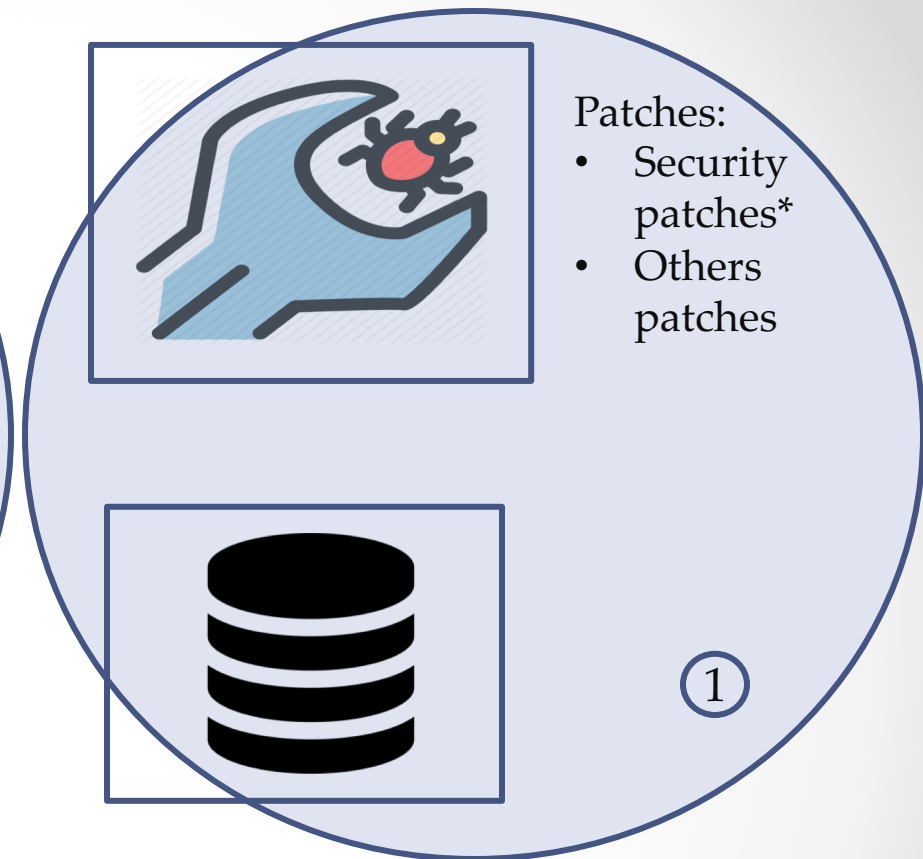
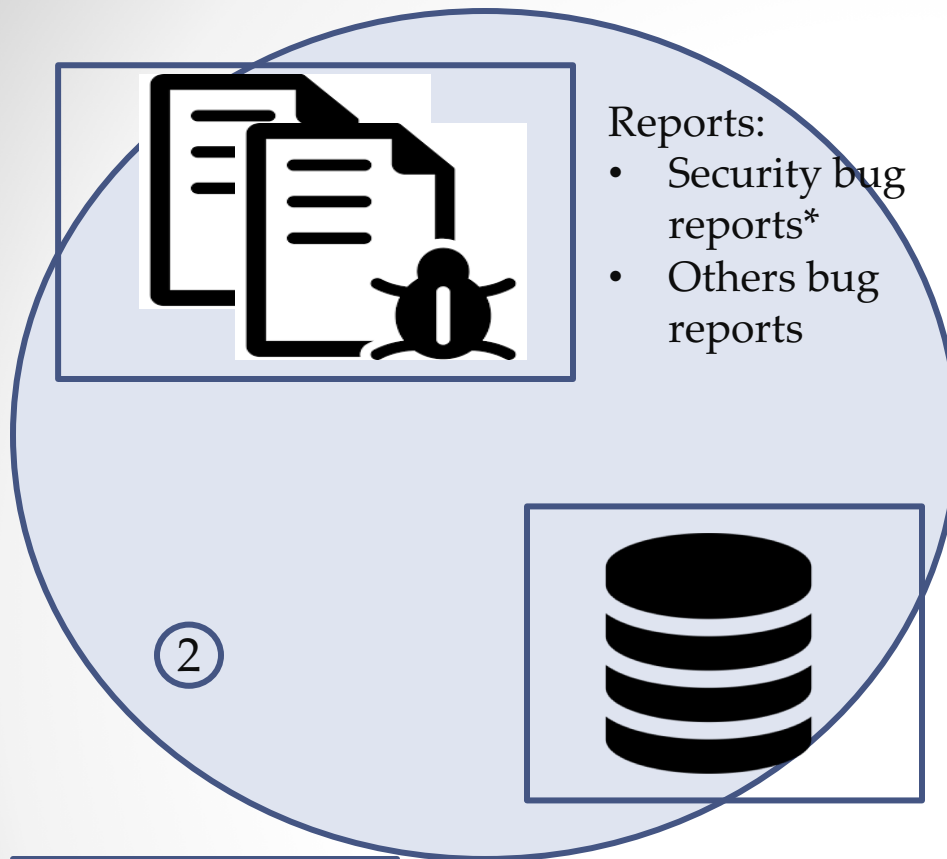
Délowendé D. A. Sawadogo (UQAM)

Naouel Moha (UQAM)

Tégawendé F. Bissyande (Univ Lu, SnT)

Second meeting of the SE@MTL community

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Bugs:

- Vulnerabilities
- Others bugs

« Silent security patches »

- Non flagged
- Suppose to be a patch without security impact

« Silent security bug repports »

- Non flagged
- Suppose to be a bug repport without security impact

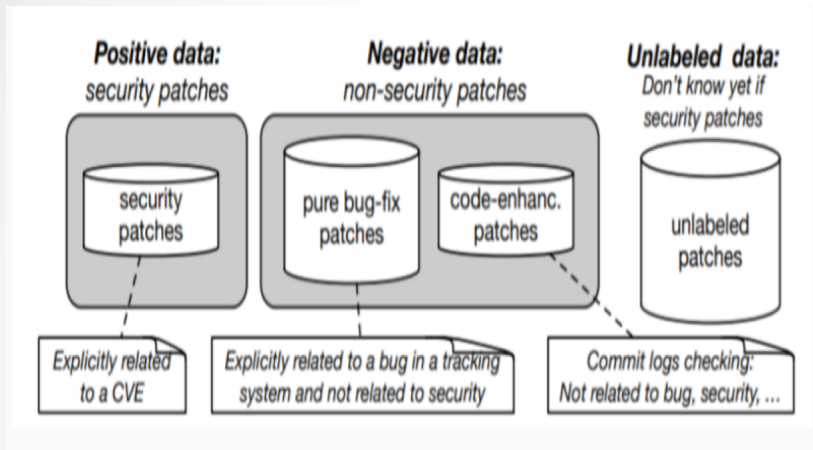
Why?

(1) Silent security patches detection

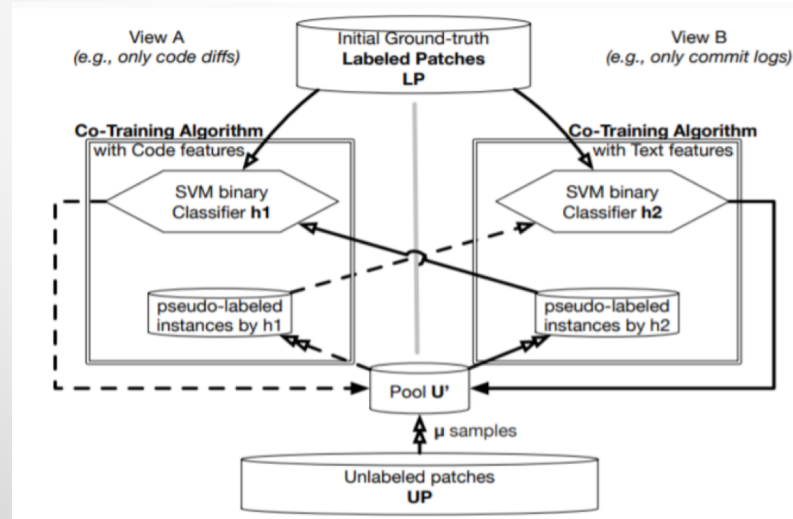
- ❖ Binary classification: we aimed to detect silent vulnerabilities in a set of fixes.
- ❖ We used text analysis on commits logs and code analysis on commits diffs.
- ❖ We trained a prediction model based on machine learning algorithms.



(1) Silent security patches detection



| ID | code-fix features | ID | security-sensitive features |
|---------|--|---------|-----------------------------------|
| F1 | #files changed in a commit | F1 | #Sizeof added |
| F2 | #Loops added | F2 | #Sizeof removed |
| F3 | #Loops removed | F3 | F1 - F2 |
| F4 | F2 - F3 | F4 | F1 + F2 |
| F5 | F2 + F3 | F5-F6 | Similar to F1 to F2 for #continue |
| F6-F9 | Similar to F2 to F5 for #ifs | F7-F8 | Similar to F1 to F2 for #break |
| F10-F13 | Similar to F2 to F5 for #Lines | F9-F10 | Similar to F1 to F2 for #INTMAX |
| F14-F17 | Similar to F2 to F5 for #Parenthesized expressions | F11-F12 | Similar to F1 to F2 for #goto |
| F18-F21 | Similar to F2 to F5 for #Boolean operators | F13-F14 | Similar to F1 to F2 for #define |
| F22-F25 | Similar to F2 to F5 for #Assignments | F15-F18 | Similar to F1 to F4 for #struct |
| F26-F29 | Similar to F2 to F5 for #Functions call | F19-F20 | Similar to F1 to F2 for #offset |
| F30-F33 | Similar to F2 to F5 for #Expression | F21-F24 | Similar to F1 to F4 for #void |
| ID | text features | | |
| W1-W10 | 10 Most recurrent non-stop words | | |



- ✓ we investigate the discriminative power of a variety of features to clarify the possibility of a learning process.
- ✓ We propose a semi-supervised approach with Co-Training which we demonstrate to yield high precision (95%) and recall (88%).
- ✓ we show that our approach can help to flag patches that were unlabeled until now.

(2) Security bug reports detection

- ❖ Mining 204 Open sources projects with arround 2000 labelled (positive and negative) vulnerabilities commits (Ponta et al., 2019) ✓

Extraction of Bug reports by links in these commits. ✓

Training an automatic learning model to identify security sensitive bugs reports.



Bibliography

- (1) Ponta et al., A Manually-Curated Dataset of Fixes to Vulnerabilities of Open-Source Software, arXiv preprint arXiv:1902.02595v (2019)