

Universidade do Minho

Engenharia de Segurança

Vulnerabilities Mapping

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Vulnerability Management

- Area of Information Security
- Management of vulnerabilities found in:
 - pentests
 - bug bounty programs
 - user contribution
 - researches

Vulnerability Management

- Various Data Source

- Nessus
- Qualys
- Acunetix
- OpenVas
- Nmap
- Nexpose
- so on

Vulnerability Representation

| | Nessus | Qualys |
|-------------------|---|---|
| ID | 10669 | 10340 |
| Name | A1Stats Multiple Script Traversal Arbitrary File Access | Drummon Miles A1Stats Directory Traversal Vulnerability |
| Categories | infos | Remote Discovery, Patch Available, Exploit Available |
| Family | CGI abuses | CGI |
| CVE | CVE-2001-0561 | CVE-2001-0561 |
| CVSS Score | Medium / CVSS Base Score : 5.0 | 7.5 |
| Bugtraq ID | 2705 | 2705 |

Problem

How to map vulnerabilities from different sources?

Solution

- Compare the attributes:
 - Title
 - CVE
 - References
- Calculate the similarity

Solution

- 2 reference datasets:
 - Known matches
 - Known unique vulnerabilities (non-matches)

Build Reference Datasets

- Matches:
 - 141 entries
 - N to N
- Not Matches:
 - 67 entries/scanner
 - 134 entries total
 - Vulns that do not have mappings

Build Test DB

- Actual DB: ~150MB (csv files) and ~220k entries
- Test DB: <1MB (csv files) and 407 entries

Comparing Attributes

- Python FuzzyWuzzy
- Title:
 - Token Sort Ratio (doesn't ignore duplicates, but ignores order)
 - Never empty
- CVE:
 - Token Set Ratio (ignores duplicates)
 - May be empty
- Refs:
 - Token Set Ratio (ignores duplicates)
 - May be empty

Calculating Similarity

$$\text{Similarity} = (\text{TitleRatio} + 2 * \text{CVERatio} + 2 * \text{RefsRatio}) / (1 + 2 * [0|1] + 2 * [0|1])$$

Example:

$$\text{Similarity} = (0.65856 + 2 * 0.72341 + 2 * 0) / (1 + 2 + 0) = 0.70179$$

[0|1] => If CVE/Refs were empty, it doesn't count

Matches Results

- True Positives:

- For each entry E on the DB, do:
 - $s = \text{similarity}(P, E)$
- if $E == Q$ and $s > 0.65$:
 - True Positive

- False Positives:

- For each entry E on the DB, do:
 - $s = \text{similarity}(P, E)$
- if $E == Q$ and $s < 0.65$:
 - False Positive

Confusion Matrix

| Predicted | Positive | Negative |
|-----------|----------|----------|
| Positive | 127 | 14 |
| Negative | 125 | 9 |

Accuracy, Precision, Recall and F_1 Score

- Accuracy = Correct Classification / Total Entries = 0.4945
- Precision = $TP / (TP + FP) = 127 / (127 + 14) = 0.9007$
- Recall = $TP / (TP + FN) = 127 / (127 + 125) = 0.5040$
- F_1 Score = $2 * (Precision * Recall) / (Precision + Recall) = 0.6463$

Future Work

- Graphs:
 - ROC (Receiver Operating Characteristic)
 - AUC (Area Under the Curve)
- Adjust Similarity Formula
- Use more attributes to compose the formula
- Run it through the actual DB (~220k entries)
- Add other sources (OpenVAS, Acunetix, etc)